

Methodical Accelerator Design

Overview of 'Next Generation'

FCCIS WP2 - CERN.

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CERN-BE/ABP

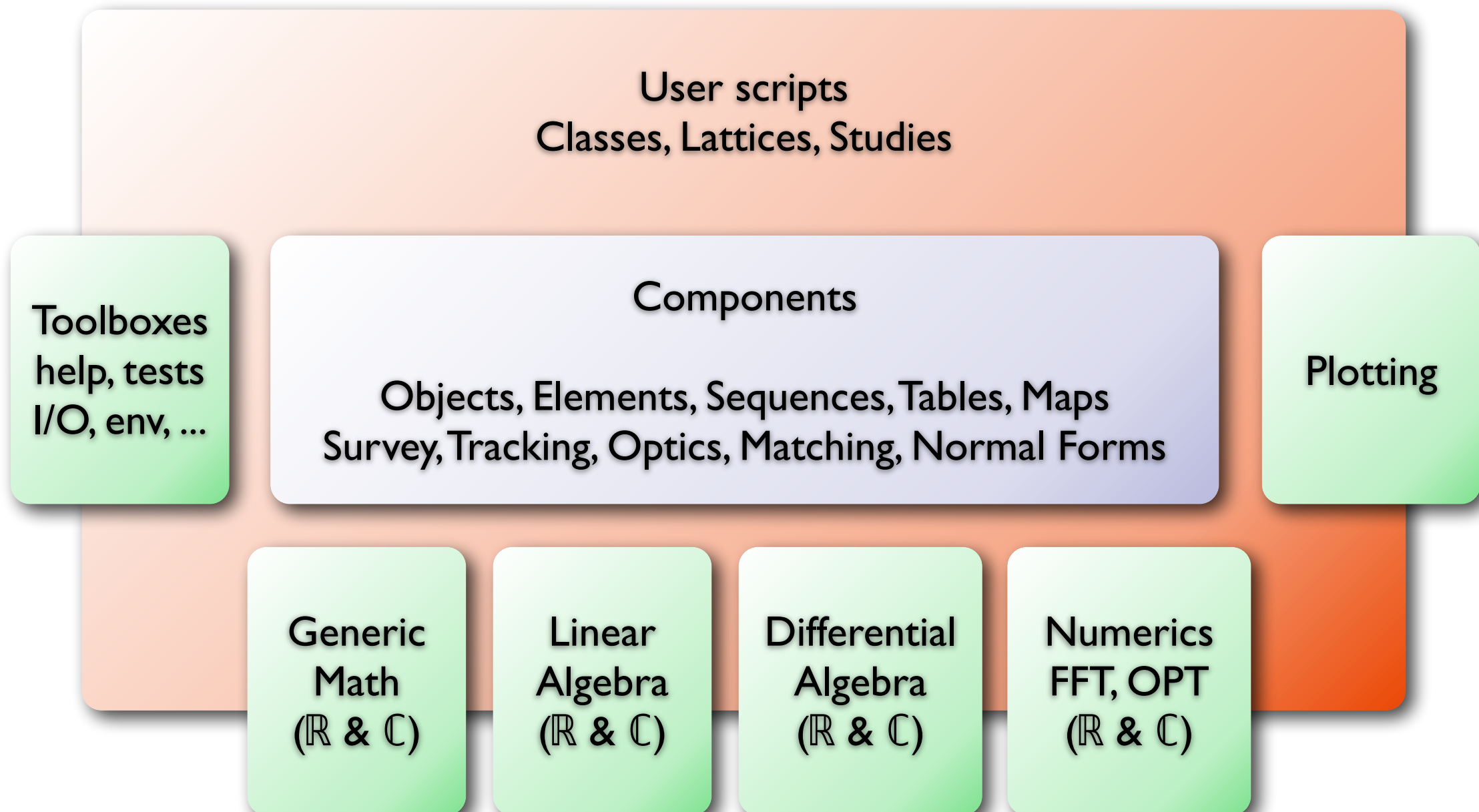
1st December 2021

- **Long term design: easy to use and extend.**
 - ➔ Flexible language ➡ **fast, simple, and general purpose scripting language.**
 - ~70% of the code is written in the scripting language, ~30% in C.
 - ➔ Flexible technologies ➡ **self-contained, all-in-one and modular.**
 - single application, no dependencies (except Gnuplot for plotting).
 - ➔ Efficient & Portable technologies ➡ **embeds a Just in Time compiler.**
 - same results everywhere (LNX, OSX, WIN), extensive unit tests (>8000).
 - fast and extremely simple Foreign Function Interface to C, C++, Fortran, etc...
- **6D PTC physics using GTPSA (for DA) and symplectic integrators.**
 - slicing, combined physics, combined elements, support/development for extensions is easy...
- **Development open source.**
 - ➔ GitHub <https://github.com/MethodicalAcceleratorDesign/MAD>
 - ➔ License GPL V3, User manual (~180p, covers <20%), Programmer Manual (29p).

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**- intended objective -
provide a general platform to
develop tracking and optics codes
for accelerator beam physics.**

- **Built from the start as a *platform* to develop & benchmark physics.**
 - *Everything is accessible, modifiable and extensible by users from scripts (e.g. even at runtime).*



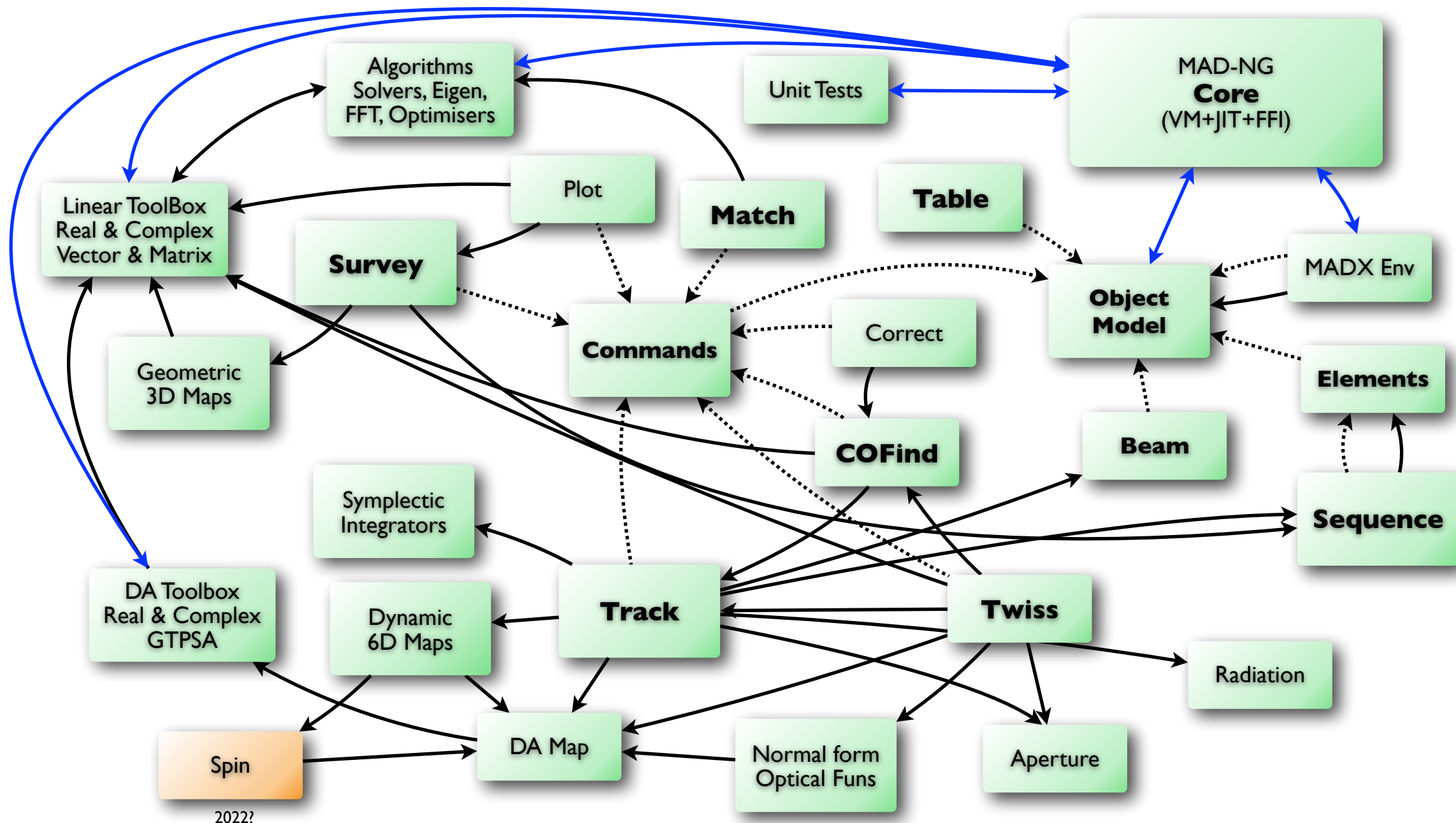
Legend

$A \xleftrightarrow{\text{exposes}} B$ $A \xrightarrow{\text{is-a}} B$ $A \xrightarrow{\text{uses}} B$
 $A \longleftrightarrow B$ $A \cdots \rightarrow B$ $A \rightarrow B$

Done

Dev

Todo

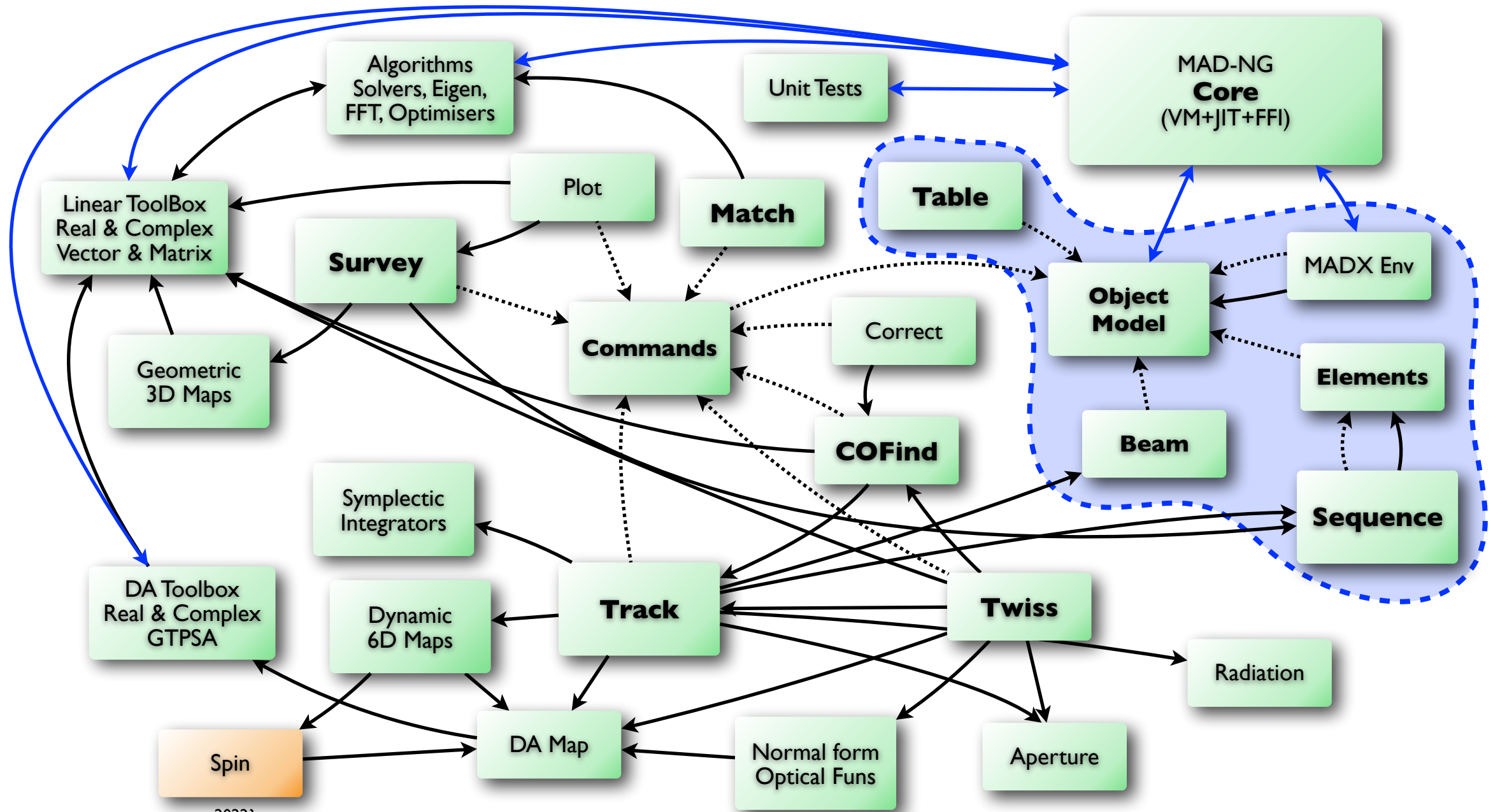


2022?

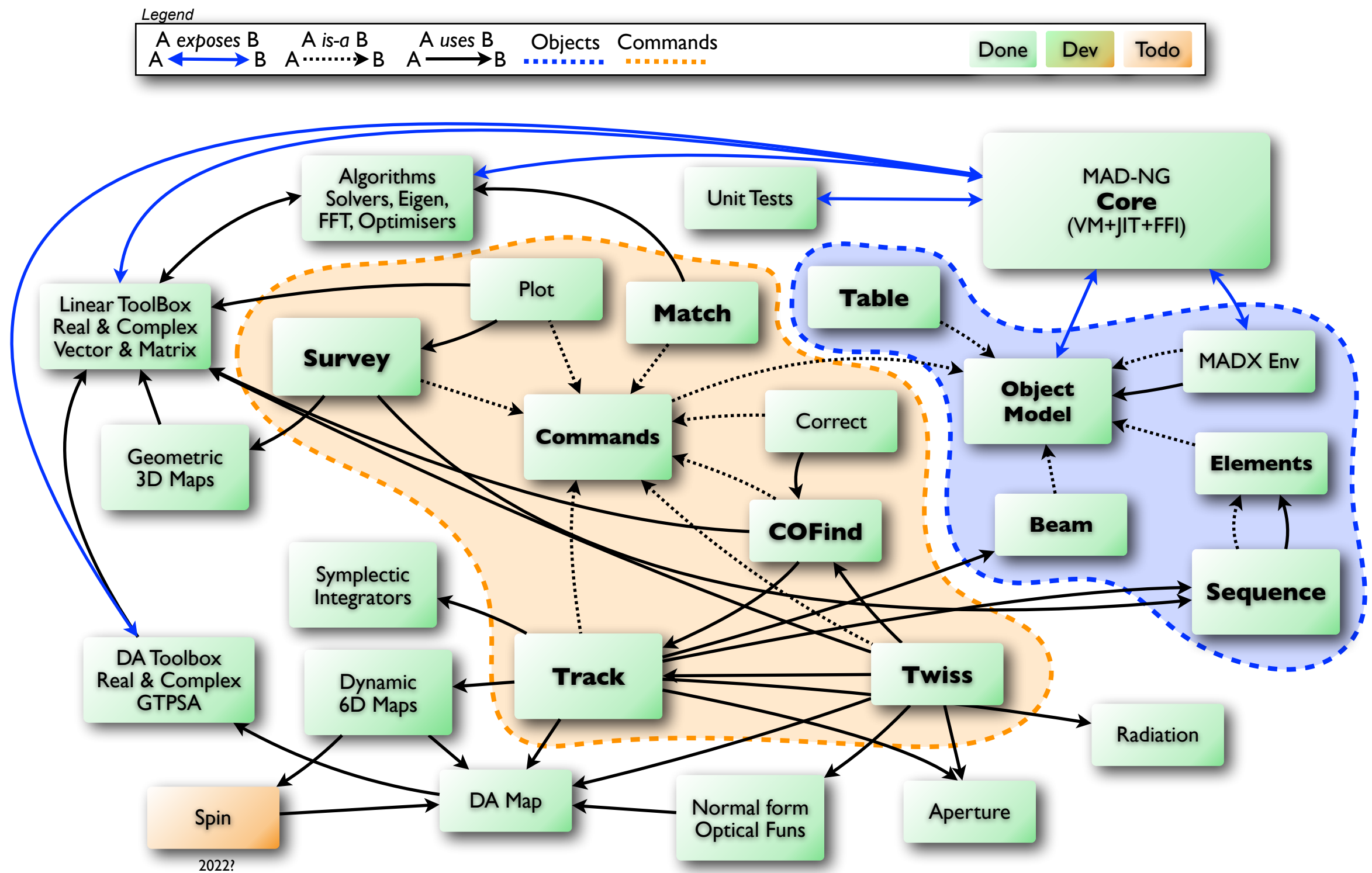
Legend

$A \xleftrightarrow{\text{blue}} B$ A exposes B
 $A \xrightarrow{\text{dotted}} B$ A is-a B
 $A \xrightarrow{\text{black}} B$ A uses B
 Objects

Done
Dev
Todo



2022?



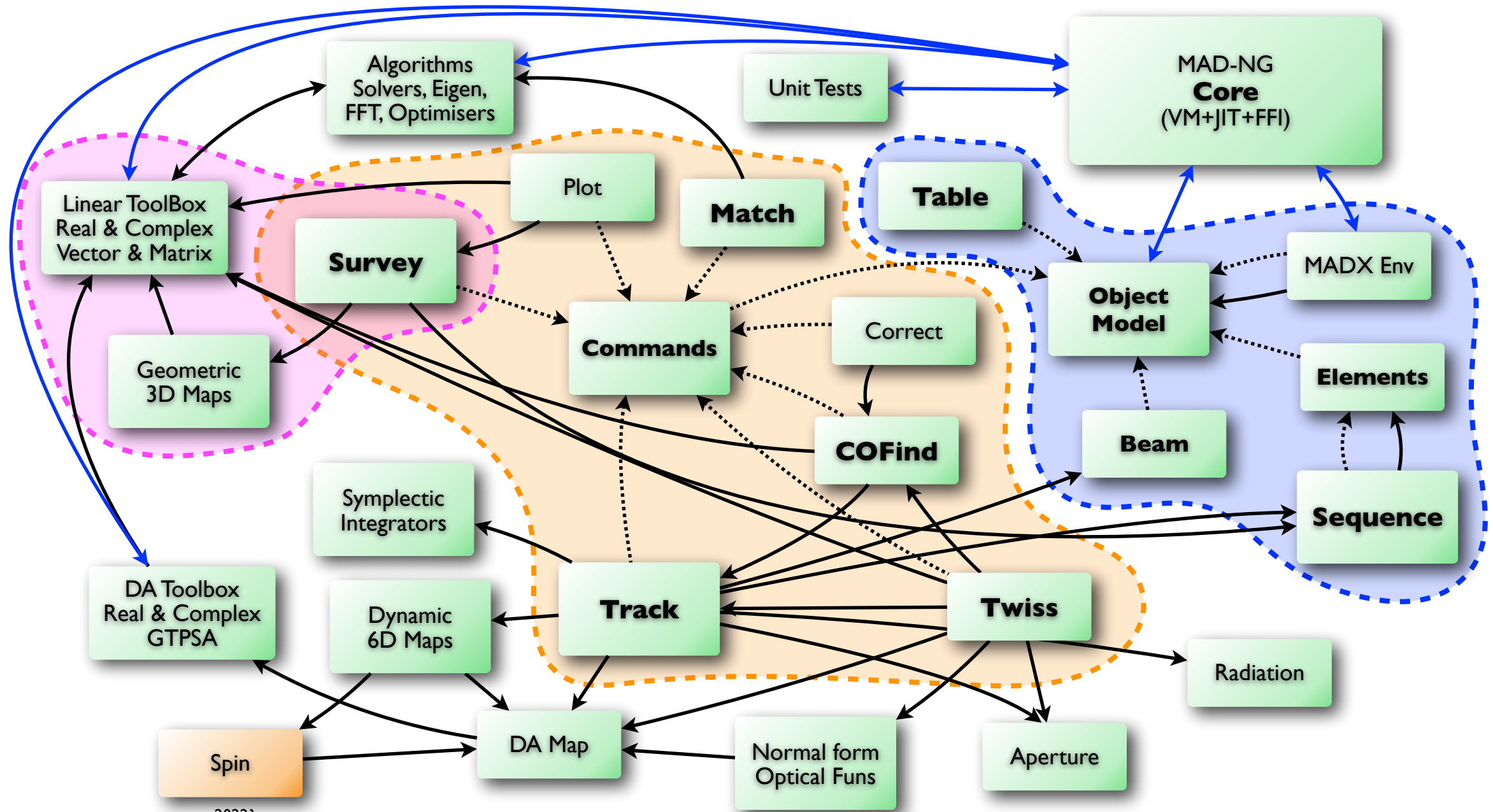
Legend

$A \xleftrightarrow{\text{blue}} B$: A exposes B
 $A \xrightarrow{\text{dotted}} B$: A is-a B
 $A \xrightarrow{\text{black}} B$: A uses B
---- : Objects
---- : Commands
---- : Geo/LinAlg

Done

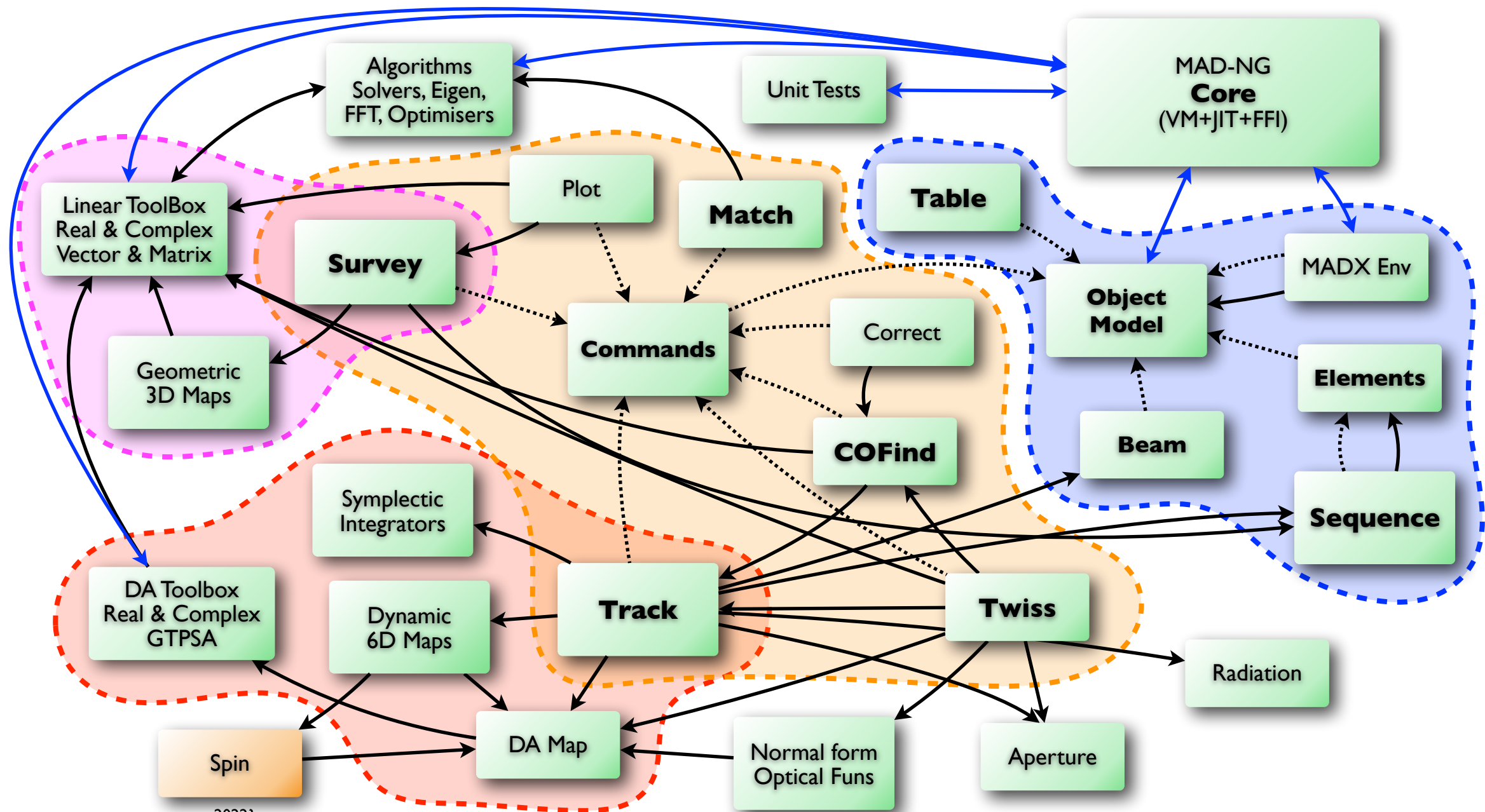
Dev

Todo



2022?

Legend



2022?

SPS in MAD-X

```

SPS:    LINE = (6*SUPER);
SUPER: LINE = (7*P44,INSERT,7*P44);
INSERT: LINE = (P24,2*P00,P42);
P00:    LINE = (QF,DL,QD,DL);
P24:    LINE = (QF,DM,2*B2,DS,PD);
P42:    LINE = (PF,QD,2*B2,DM,DS);
P44:    LINE = (PF,PD);
PD:     LINE = (QD,2*B2,2*B1,DS);
PF:     LINE = (QF,2*B1,2*B2,DS);
    
```

SPS in MAD-NG

```

pf      = bline {qf,2*b1,2*b2,ds}
pd      = bline {qd,2*b2,2*b1,ds}
p24     = bline {qf,dm,2*b2,ds,pd}
p42     = bline {pf,qd,2*b2,dm,ds}
p00     = bline {qf,dl,qd,dl}
p44     = bline {pf,pd}
insert  = bline {p24,2*p00,p42}
super   = bline {7*p44,insert,7*p44}
SPS     = sequence 'SPS' {6*super}
    
```

© **Lattices definition as in MAD-X** (*syntax is very close*)

- ➔ sequences are both containers (e.g. access elements) and table (store arbitrary objects).
 - e.g. to store their **beam** or their own list of **knobs**.
- ➔ elements are both containers (e.g. access attributes) and table (store arbitrary objects).
- ➔ sequence can include sub**sequences**, beam **lines** and **elements** (and sub**elements**).
- ➔ operator overloading (+, -, *) allows to mix *lines* and *sequences* descriptions arbitrarily.
- ➔ names are optional and can be non-unique with support for *relative* or *absolute* counts.
 - positions 'AT' can be absolute or relative 'FROM' names with absolute or relative counts.

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**unified definitions of
lines and sequences
plus extensions**

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**unified definitions of
lines and sequences
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● Manage arbitrary number of sequences to allow simulation of **entire accelerators complex**.

- ➔ Shared sequences, e.g. LHCB1 and LHCB2.
 - ▶ provides few sharing policies.
- ➔ Chained sequences, e.g. PSB, PS, SPS and BTL.
- ➔ Conditionally chained sequences (e.g. RaceTrack).
 - ▶ e.g. Booster → Ring1 if energy > 175 GeV
 - ▶ based on special *s-link* element
 - ▶ connections and conditions are performed through an arbitrary user-defined function.

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Sequences conversion (MAD-X to MAD)

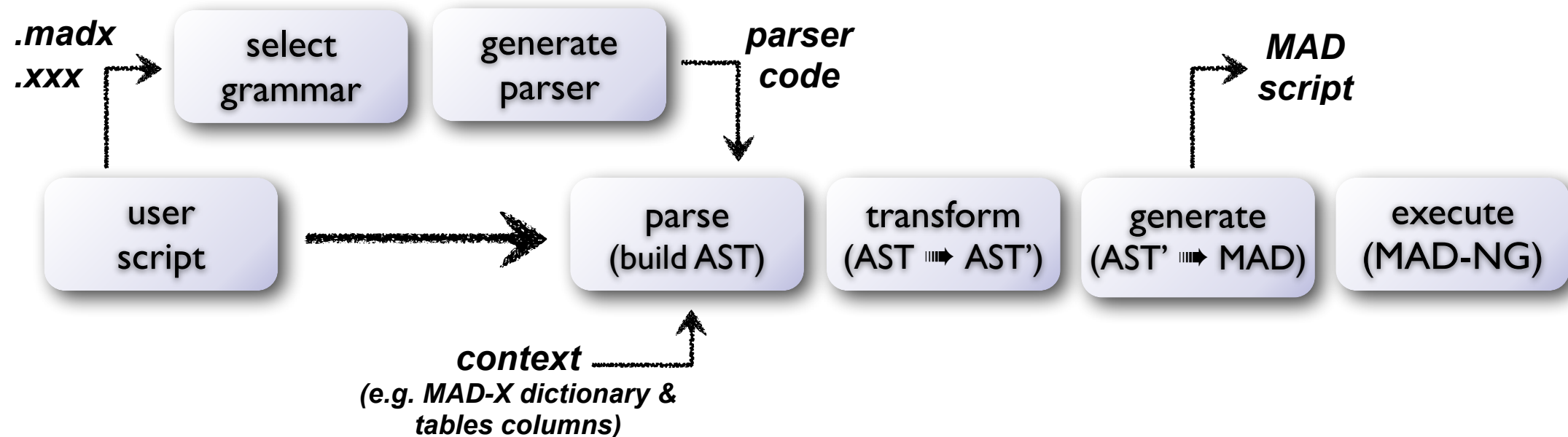
- MAD-NG **loads and convert** MAD-X sequences, elements and variables, *including deferred expressions*, **on-the-fly** into the MADX environment (a MAD-NG context that emulates MAD-X global workspace) and/or save conversion to files.

```
! convert MAD-X files on need, save to MAD file (disk), load to MADX environment (memory)
MADX:load('lhc_as-built.seq'           , 'lhc_as-built.mad')
MADX:load('opticsfile.22_ctpps2'       , 'opticsfile.22_ctpps2.mad')
MADX:load("FCCee_z_213_nosol_18.seq", "FCCee_z_213_nosol_18.mad")
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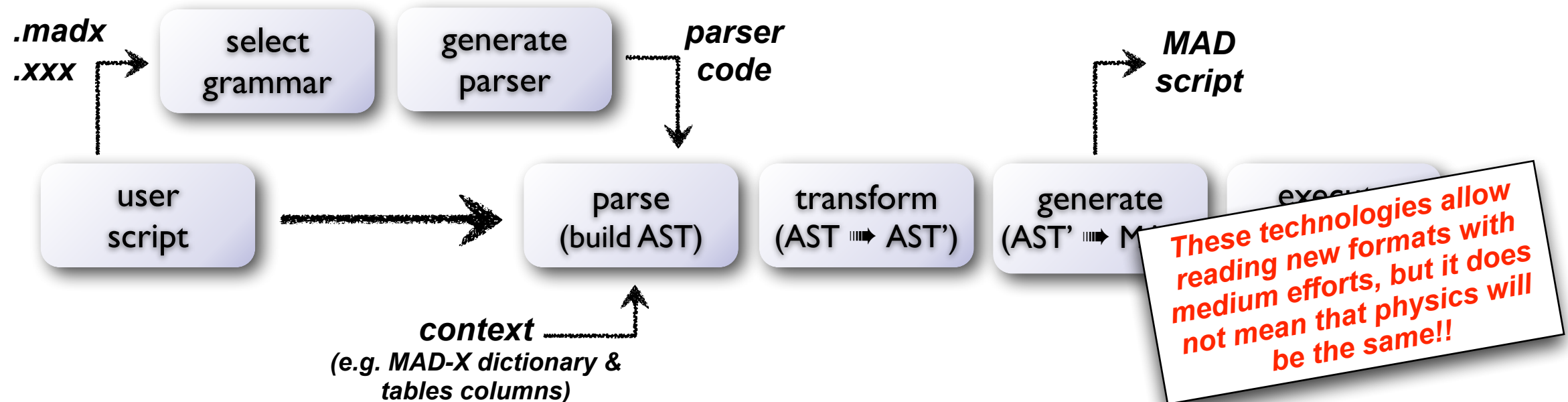
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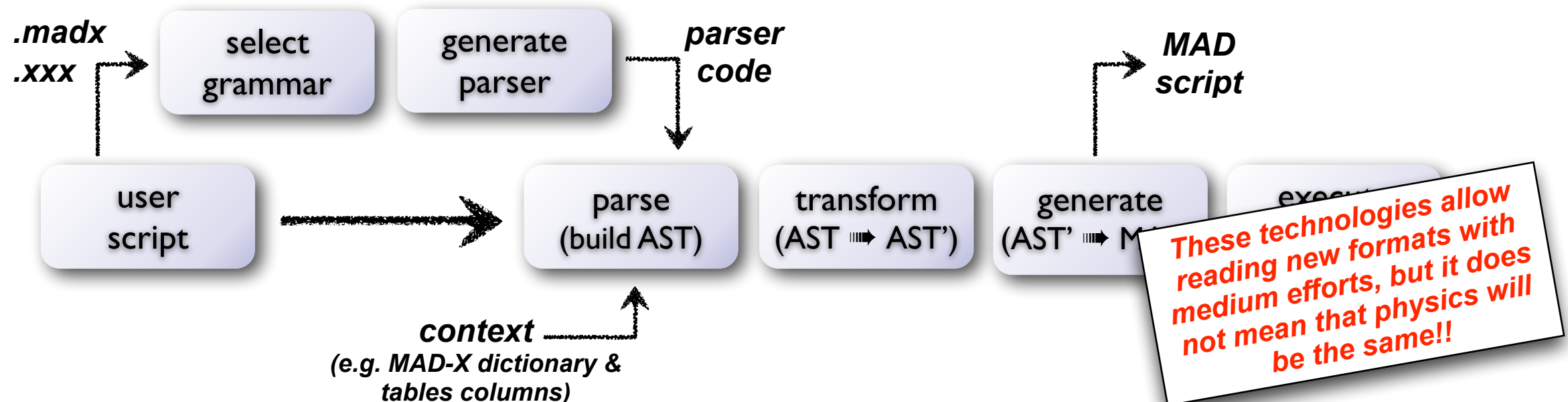
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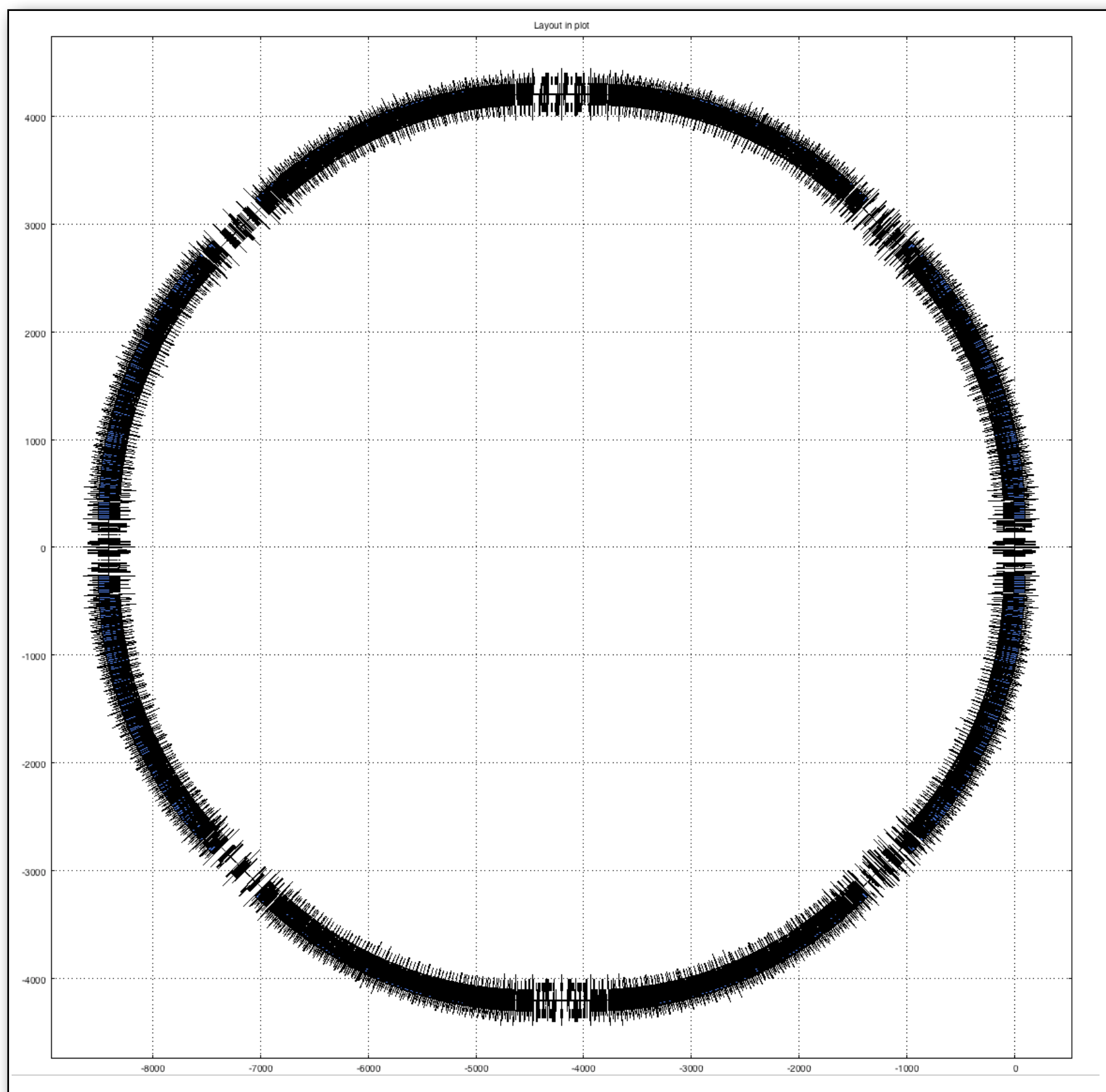


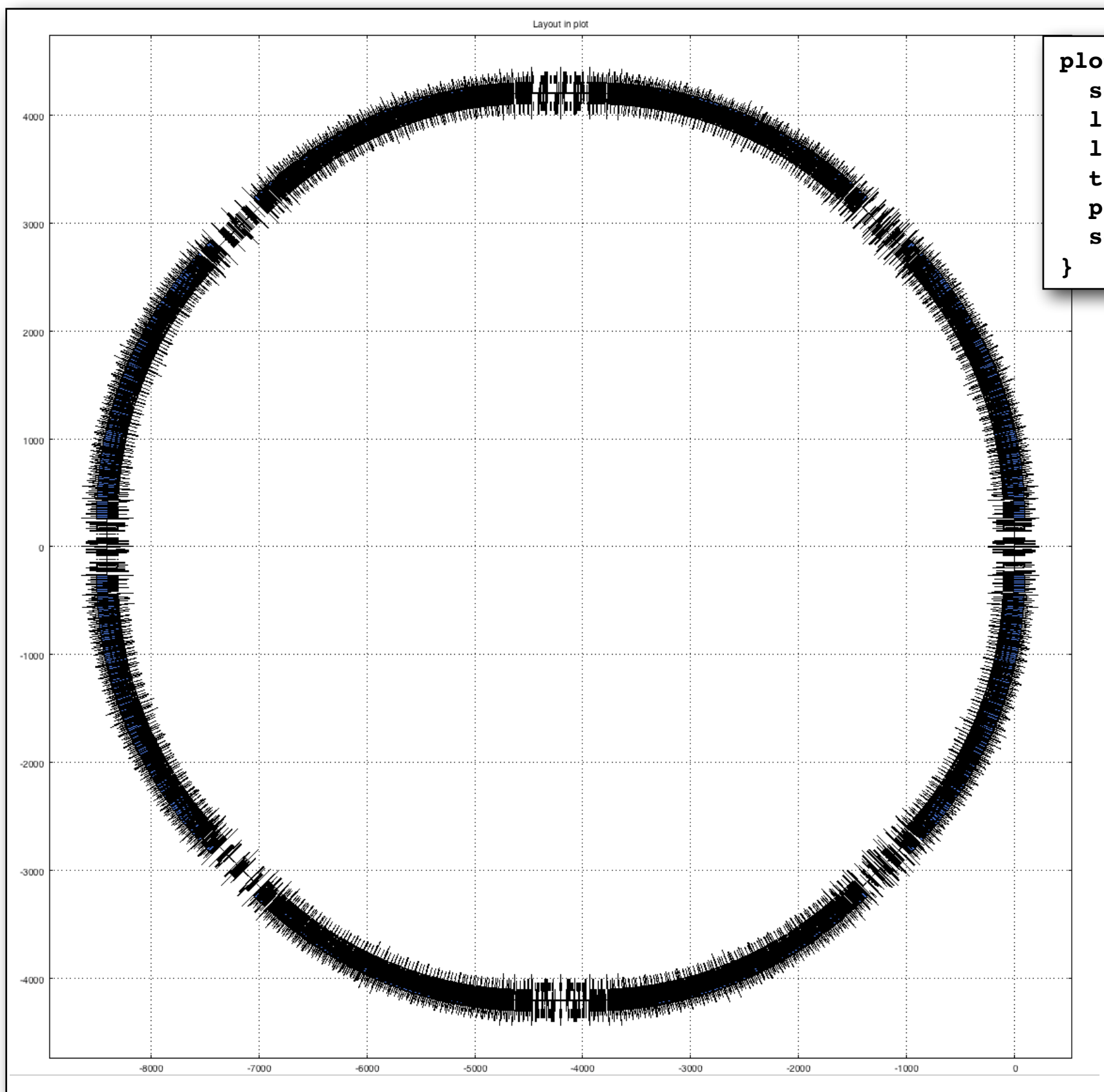
- MAD-NG allows to run MAD-X as a module to convert sequences, elements and variables into MADX environment as with CpyMad. But this method does not propagate the deferred expressions, i.e. lattice logic is lost (fine for a “static” description). Could be propagated with some extra work.



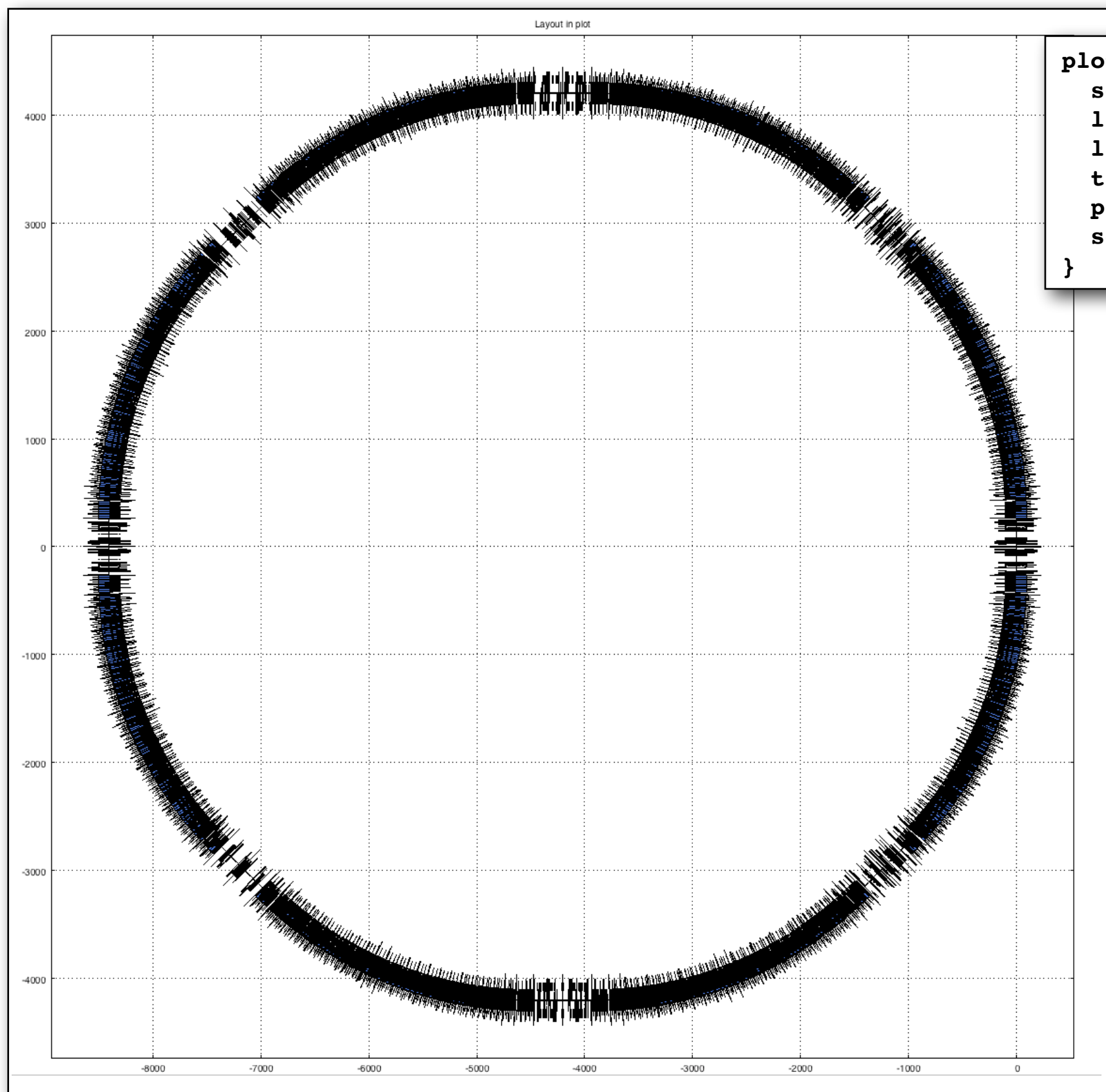
Sequence plot (*LHC 1 & 2 survey*)

Sequence plot (*LHC 1 & 2 survey*)





```
plot {
  sequence = {lhcb1, lhcb2},
  laypos   = "in",
  layonly  = false,
  title    = "Layout in plot",
  prolog   = 'set size ratio -1',
  screump  = "plotlhcb.gp",
}
```



```
plot {
  sequence = {lhcb1, lhcb2},
  laypos   = "in",
  layonly  = false,
  title    = "Layout in plot",
  prolog   = 'set size ratio -1',
  scrdump  = "plotlhcb.gp",
}
```

MAD-X loads the entire LHC definition in ~1 s.
(2 beamlines, ~30000 lines)

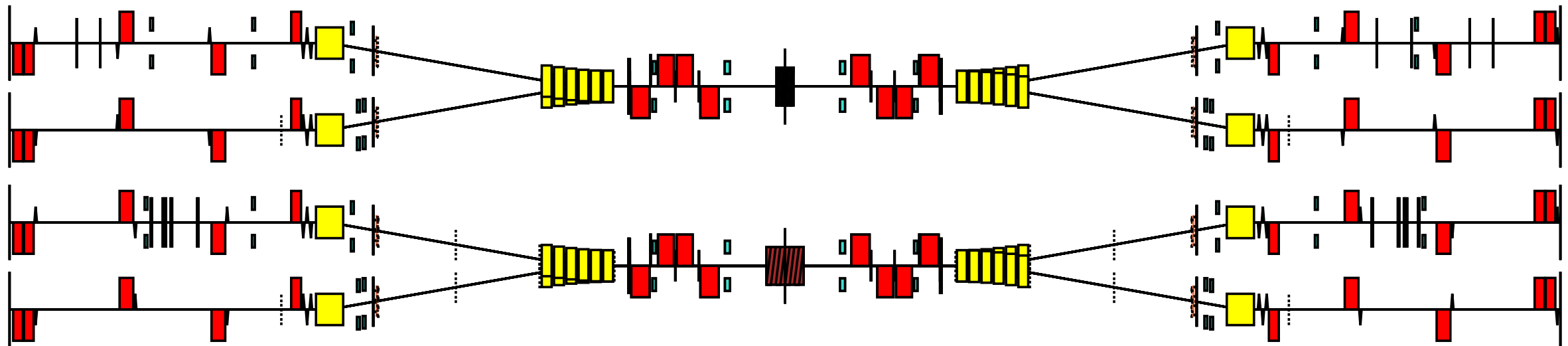
MAD-NG loads the entire LHC in MAD-X format and saved it in files in ~1 s.

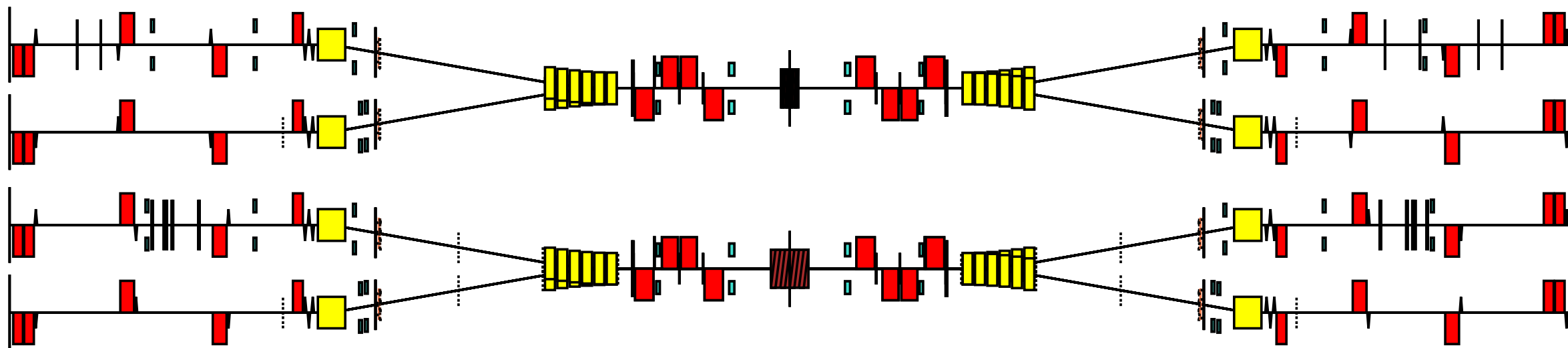
MAD-NG loads the entire LHC from converted files (.mad files) in ~0.2 s.

Gnuplot script (.gp files) size is 5 MB & 125000+ lines and takes ~1 sec to display.

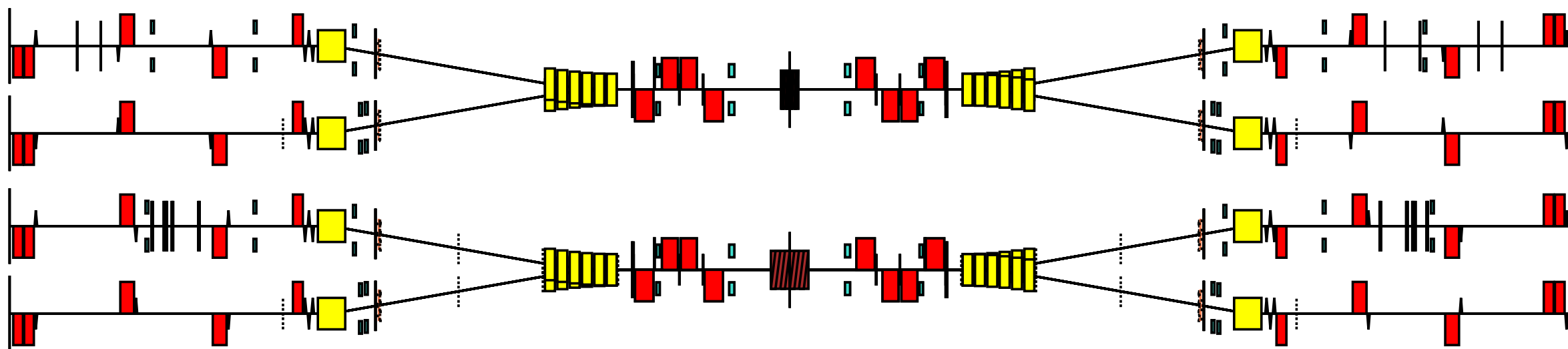
All items are tagged i.e. moving the mouse over show their name and kind

Sequence plot (*LHC 1 & 2 at IP1 & IP5 layout*)





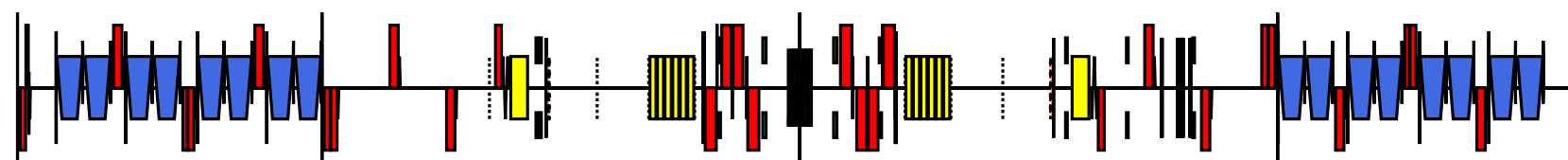
```
plot {
  sequence = { lhcb1, lhcb2, lhcb1, lhcb2 },
  range    = {
    {"E.DS.L1.B1", "S.DS.R1.B1"}, {"E.DS.L1.B2", "S.DS.R1.B2"},
    {"E.DS.L5.B1", "S.DS.R5.B1"}, {"E.DS.L5.B2", "S.DS.R5.B2"},
  },
  laydisty = {
    lhcb2["E.DS.L1.B2"].mech_sep,      ! second bline
    -0.4,                             ! third bline
    -0.4 + lhcb2['E.DS.L5.B2'].mech_sep ! fourth bline
  },
  title = "IP1-IP5 two angled beams",
}
```

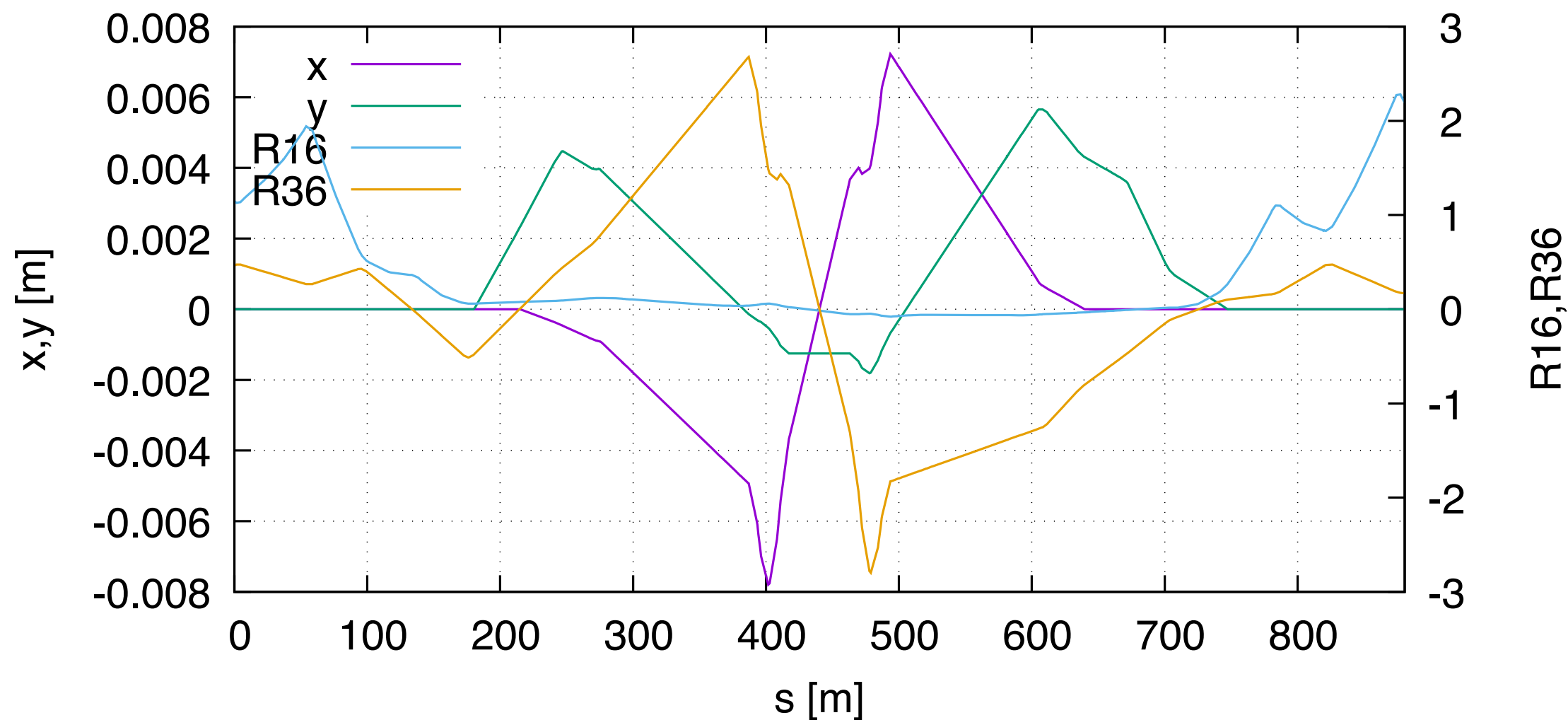
```
plot {
  sequence = { lhcb1, lhcb2, lhcb1, lhcb2 },
  range    = {
    {"E.DS.L1.B1", "S.DS.R1.B1"}, {"E.DS.L1.B2", "S.DS.R1.B2"},
    {"E.DS.L5.B1", "S.DS.R5.B1"}, {"E.DS.L5.B2", "S.DS.R5.B2"},
  },
  laydisty = {
    lhcb2["E.DS.L1.B2"].mech_sep,      ! second bline
    -0.4,                               ! third bline
    -0.4 + lhcb2['E.DS.L5.B2'].mech_sep ! fourth bline
  },
  title = "IP1-IP5 two angled beams",
}
```

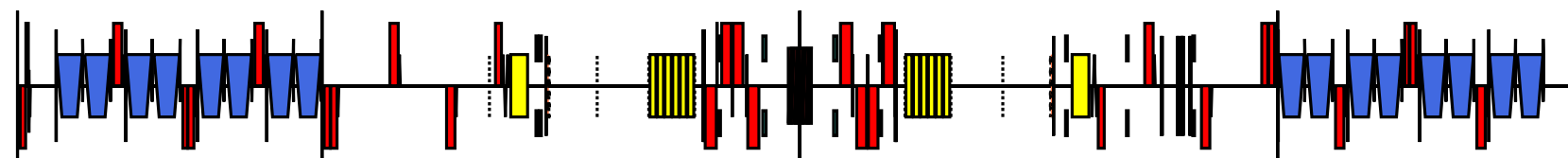


Track plot (LHCB1 around IP5)



LHCb1 around IP5



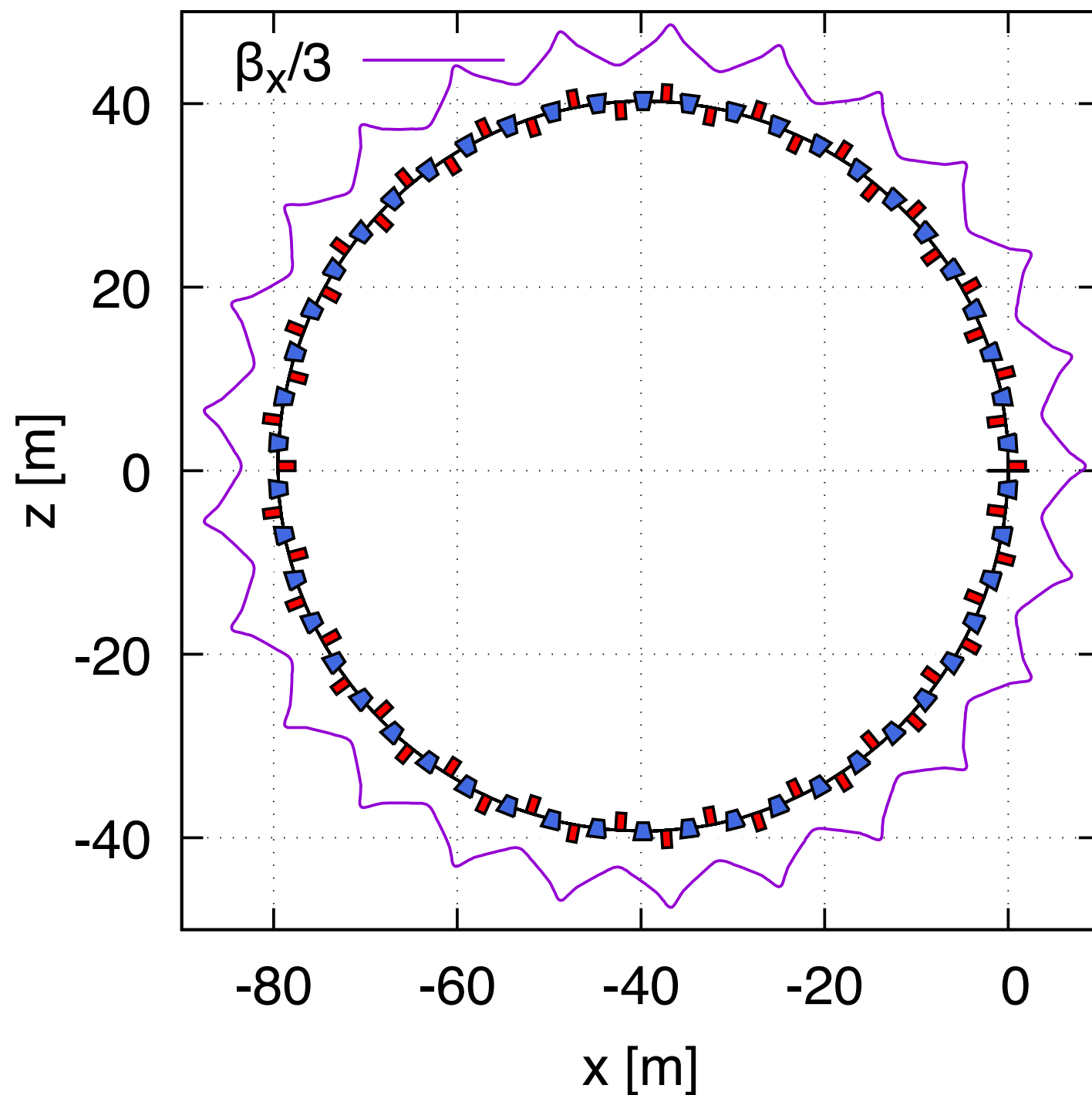


LHCB1 around IP5

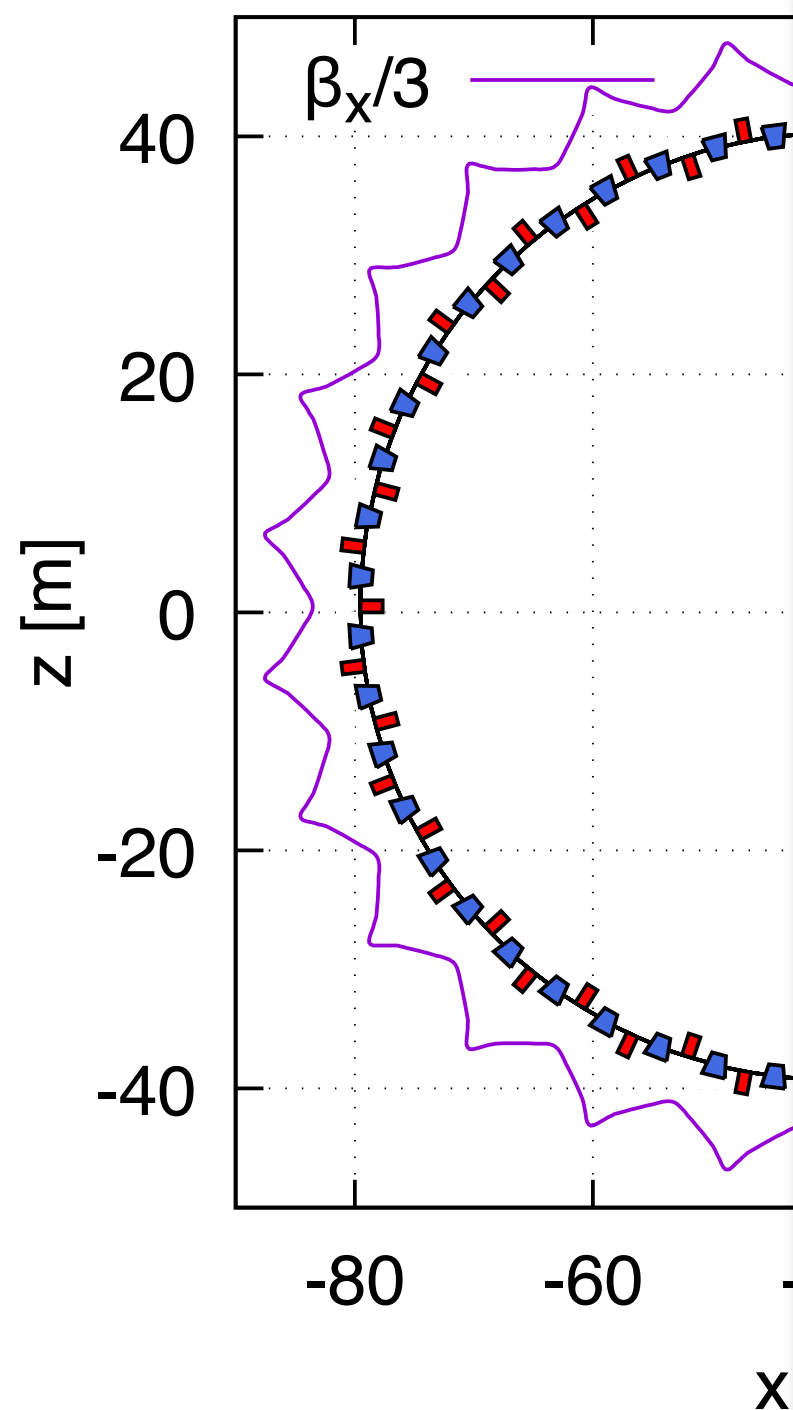


```
local e1, e2 = "E.ARC.45.B1", "S.ARC.56.B1"
mtbl:addcol('s5', \i -> mtbl.s[i]-mtbl[e1].s)
plot { -- plot with extracted data around IP5
  title      = "LHCB1 around IP5",
  sequence   = lhcb1,
  range      = {e1,e2},
  table      = mtbl,
  tablerange = {e1,e2},
  x1y1       = {s5={'x','y'}},
  x1y2       = {s5={'R16','R36'}},
  styles     = "lines",
  xlabel     = "s [m]",
  ylabel     = "x,y [m]",
  y2label    = "R16,R36",
  fontsize   = 14,
  output     = "plots/orbit_lhcb1_ip5_da.pdf",
  --screendump = "plots/orbit_lhcb1_ip5.gp",
}
```

Layout in plot with β_x



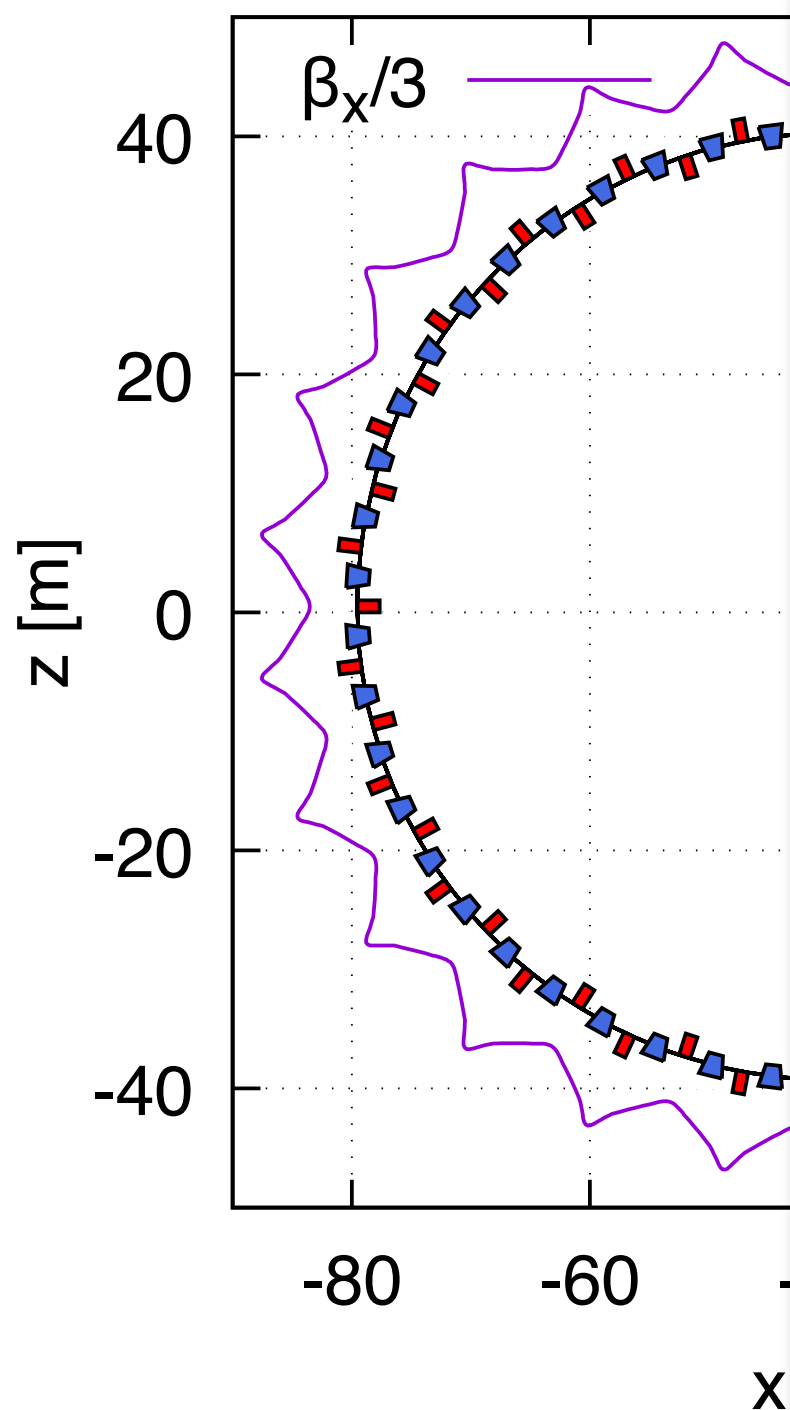
Layout in plot with β_x



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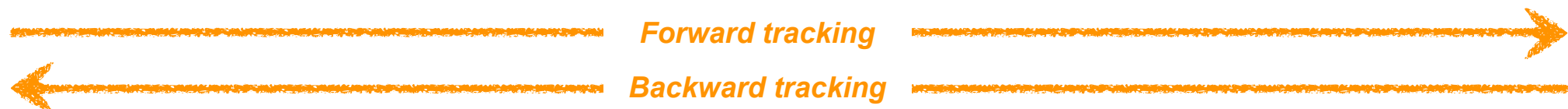
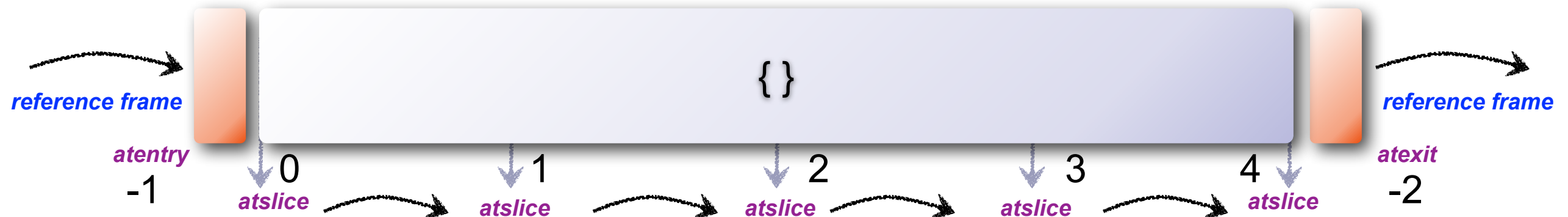
local ncell = 25
local mb = sbend      { l=2 }
local mq = quadrupole { l=1 }
local cell = sequence { l=10, refer='entry',
    mq 'mq1' { at=0, k1=0.29601 },
    mb 'mb1' { at=2, angle := pi/ncell },
    mq 'mq2' { at=5, k1=-0.30242 },
    mb 'mb2' { at=7, angle := pi/ncell },
}
local seq = sequence 'seq' { ncell*cell, beam=beam }
local sv = survey { sequence=seq, nslice=5, atslice=ftrue, mapsave=true }
local tw = twiss { sequence=seq, nslice=5, atslice=ftrue }
! compute betx in global frame
local bet11 = { x=vector(#sv), z=vector(#sv) }
local v, scl = vector(3), round(tw.bet11:max()/5)
for i=1,#sv do
    v = sv.W[i] * v:fill{3+tw.bet11[i]/scl, 0, 0}
    bet11.x[i], bet11.z[i] = v[1], v[3]
end
bet11.x = bet11.x+sv.x
bet11.z = bet11.z+sv.z
! plot layout of the ring and the betx
plot {
    sequence = seq,
    laypos   = "in",
    layonly  = false,
    title    = "Layout in plot with \u{03b2}_x",
    data     = { x=bet11.x, z=bet11.z },
    x1y1     = { x = 'z' },
    styles   = 'lines',
    xlabel   = "x [m]",
    ylabel   = "z [m]",
    legend   = { z = '\u{03b2}_x/'..scl },
}
    
```

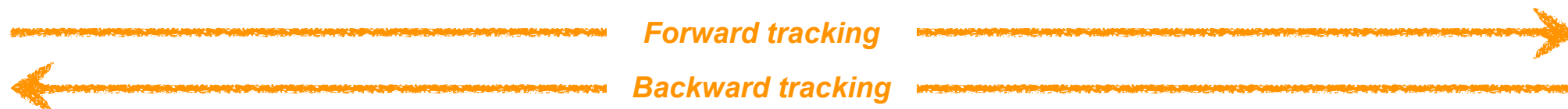
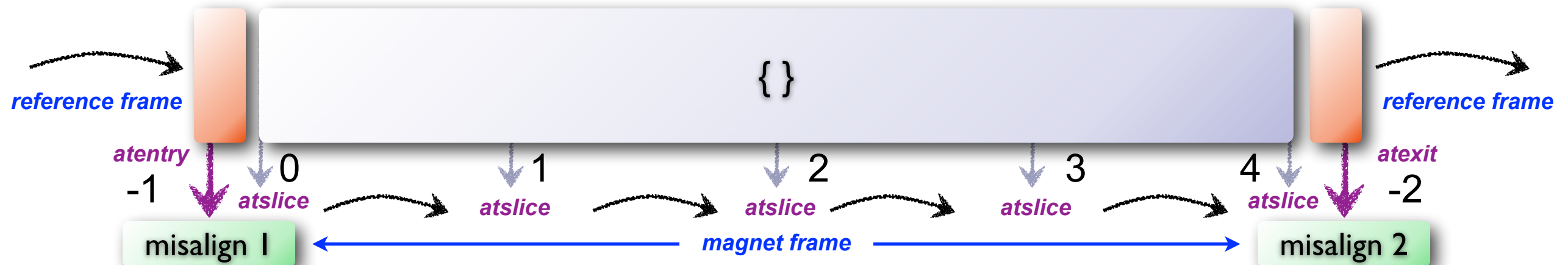
Layout in plot with β_x

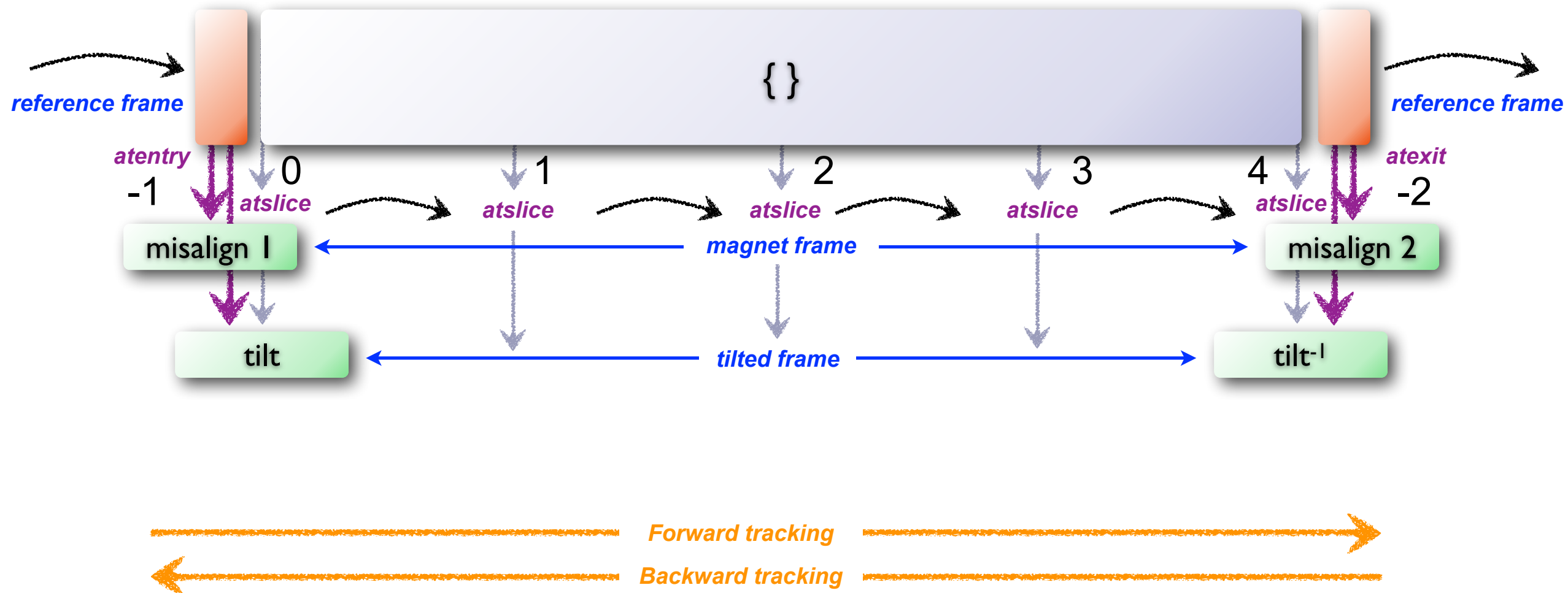


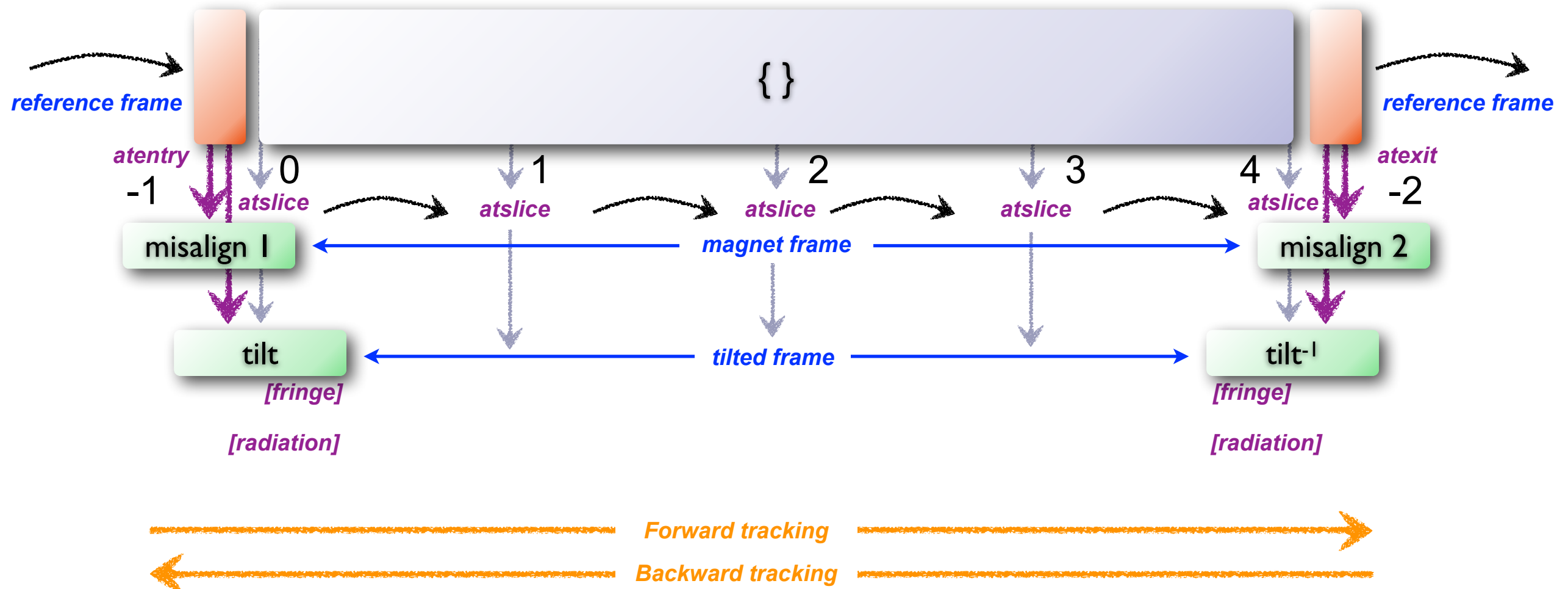
```

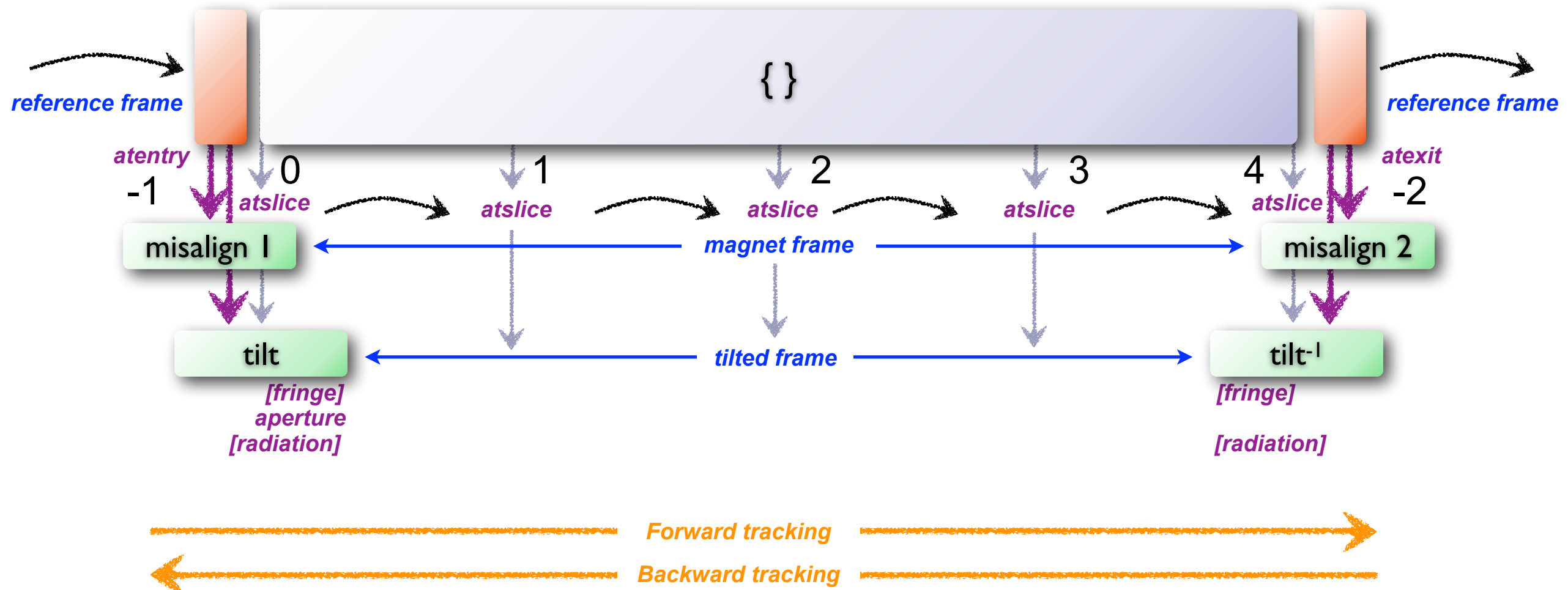
local ncell = 25
local mb = sbend      { l=2 }
local mq = quadrupole { l=1 }
local cell = sequence { l=10, refer='entry',
    mq 'mq1' { at=0, k1=0.29601 },
    mb 'mb1' { at=2, angle := pi/ncell },
    mq 'mq2' { at=5, k1=-0.30242 },
    mb 'mb2' { at=7, angle := pi/ncell },
}
local seq = sequence 'seq' { ncell*cell, beam=beam }
local sv = survey { sequence=seq, nslice=5, atslice=ftrue, mapsave=true }
local tw = twiss { sequence=seq, nslice=5, atslice=ftrue }
! compute betx in global frame
local bet11 = { x=vector(#sv), z=vector(#sv) }
local v, scl = vector(3), round(tw.bet11:max()/5)
for i=1,#sv do
    v = sv.W[i] * v:fill{3+tw.bet11[i]/scl, 0, 0}
    bet11.x[i], bet11.z[i] = v[1], v[3]
end
bet11.x = bet11.x+sv.x
bet11.z = bet11.z+sv.z
! plot layout of the ring and the betx
plot {
    sequence = seq,
    laypos    = "in",
    layonly   = false,
    title     = "Layout in plot with \u{03b2}_x",
    data      = { x=bet11.x, z=bet11.z },
    x1y1      = { x = 'z' },
    styles    = 'lines',
    xlabel    = "x [m]",
    ylabel    = "z [m]",
    legend    = { z = '\u{03b2}_x/' .. scl },
}
    
```

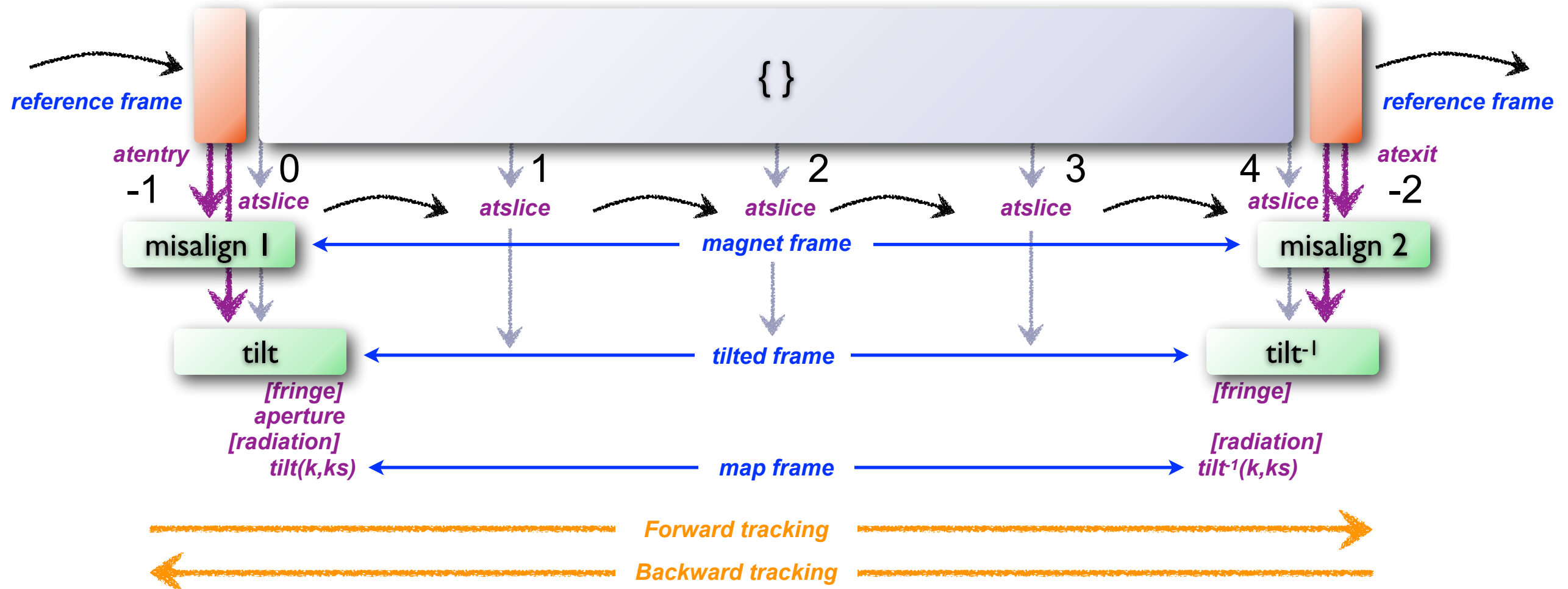


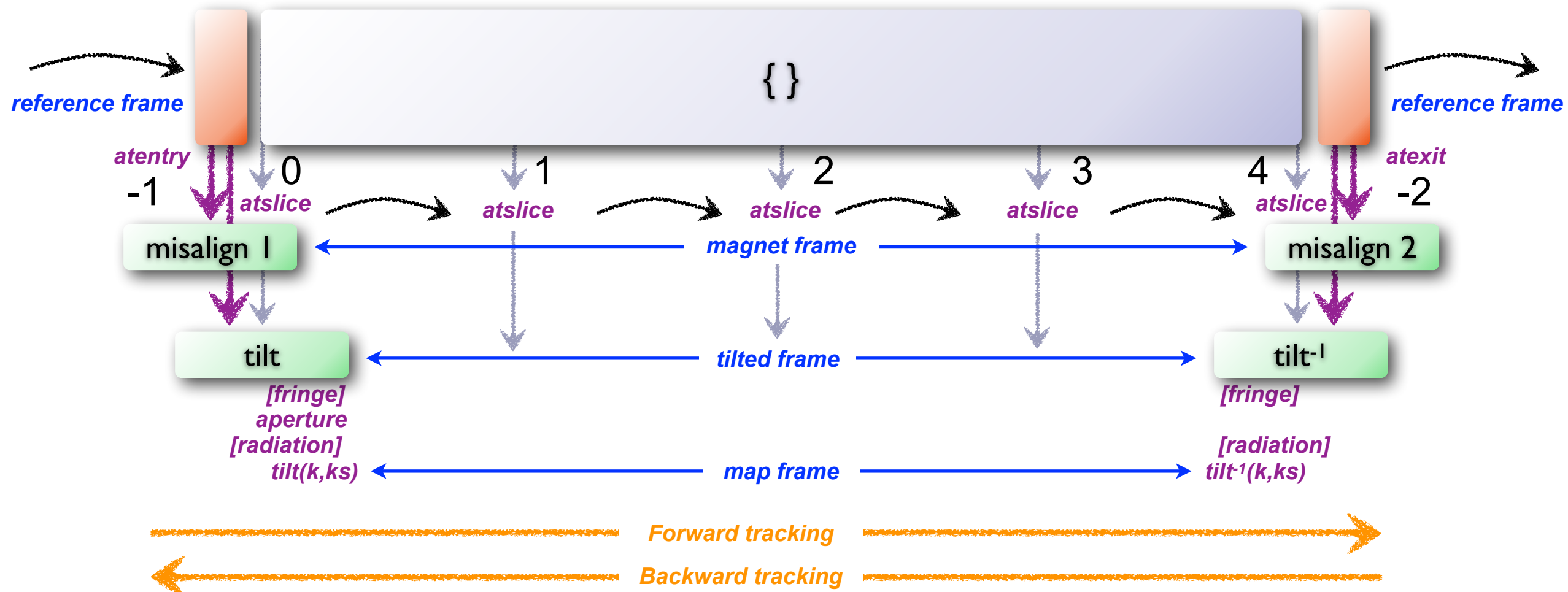






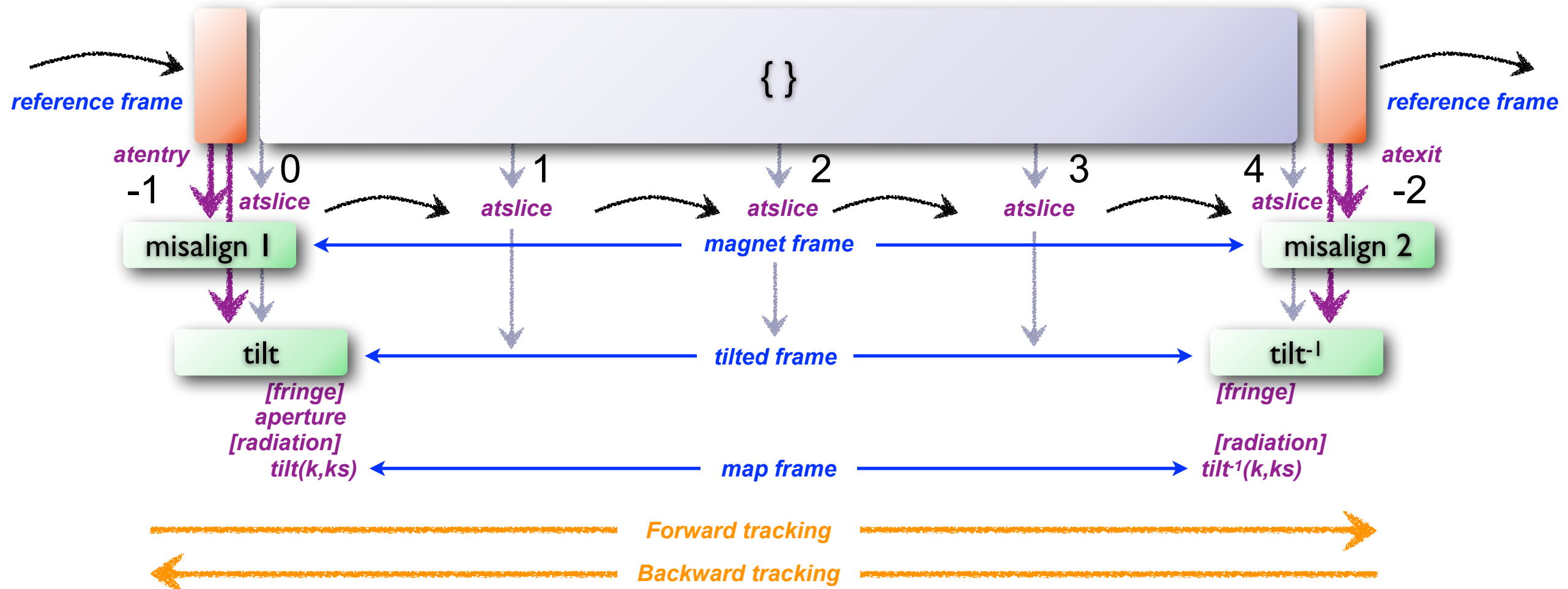






```

atentry(elm, m, sdir, -1)
mis      (elm, m, sdir)
rot      (tlt, m, sdir)
fringe   (elm, m, sdir)
track    (elm, m, 1, thick, thin)
fringe   (elm, m, -sdir)
rot      (tlt, m, -sdir)
mis      (elm, m, -sdir)
atexit   (elm, m, -sdir, -2)
    
```

- **Slicing** can be uniform or arbitrary.
- **Subelements** thick or thin can be inserted at arbitrary relative (to parent length) or absolute (from parent entry) positions. Subelements define slices.
- **Installing** elements in sequence automatically (user-policy) insert them as subelement upon collision.
- **Misalignments** (element to sequence) restore the frame on exit. Permanent misalignments (element property) don't (i.e. **patches**). Survey can consider misalignments (user-policy) for superposition inside elements.

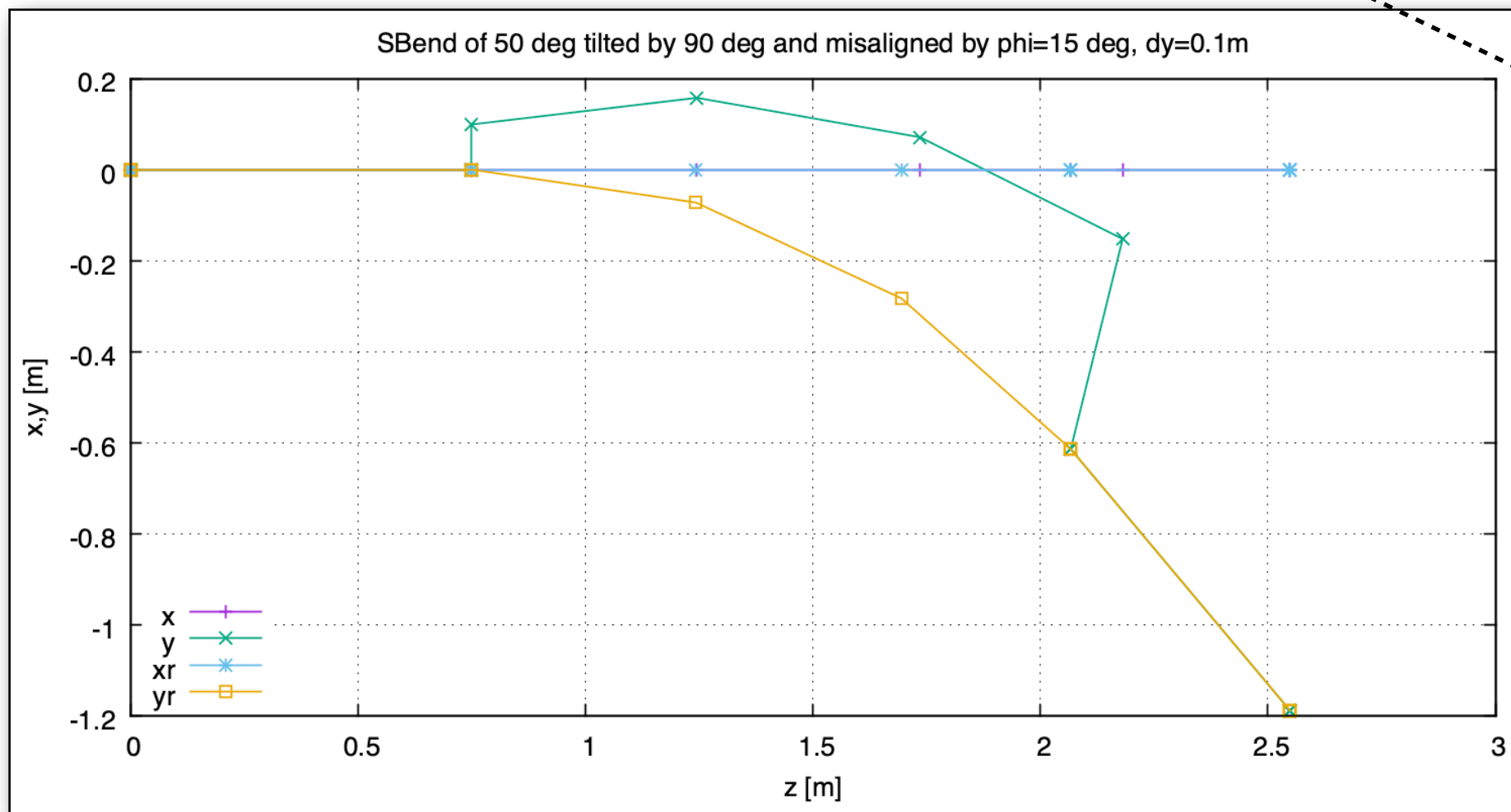
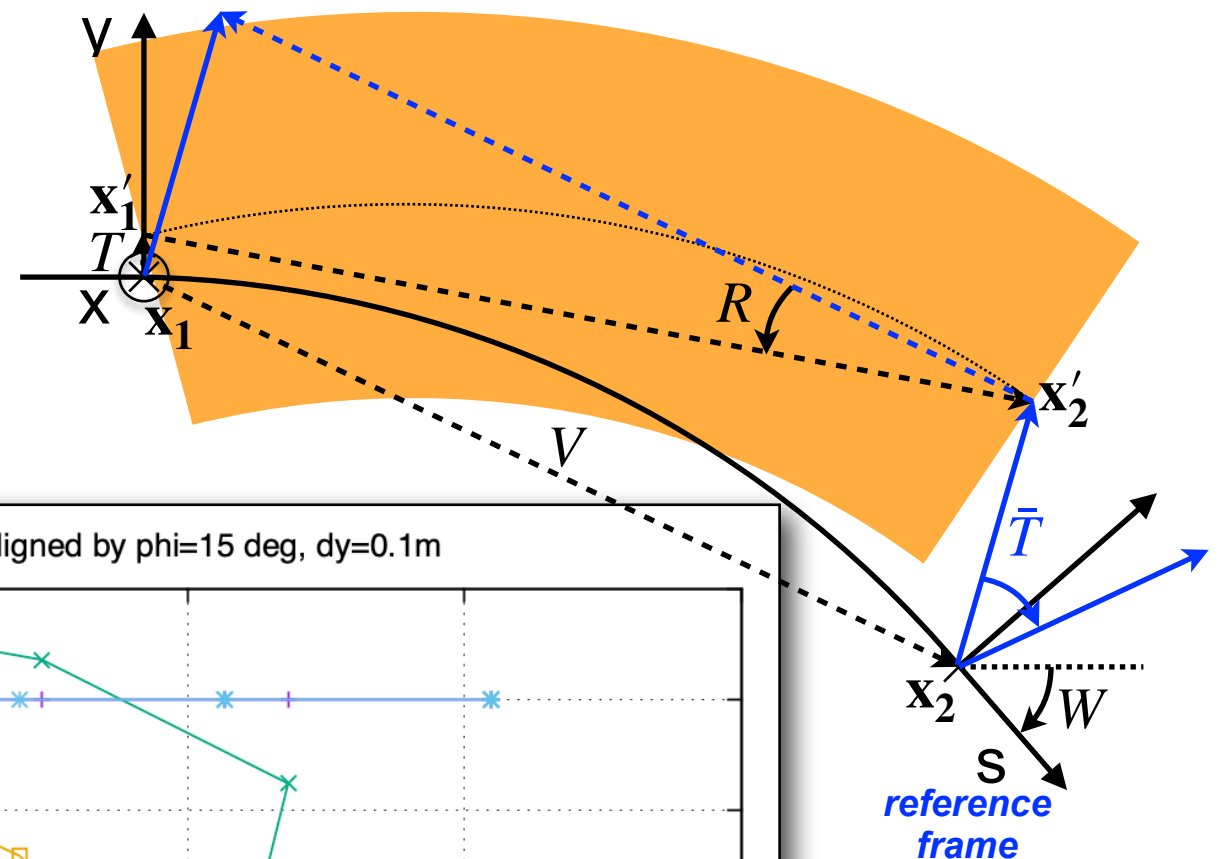
```

atentry(elm, m, sdir, -1)
mis      (elm, m, sdir)
rot      (tlt, m, sdir)
fringe   (elm, m, sdir)
track    (elm, m, 1, thick, thin)
fringe   (elm, m, -sdir)
rot      (tlt, m, -sdir)
mis      (elm, m, -sdir)
atexit   (elm, m, -sdir, -2)
    
```



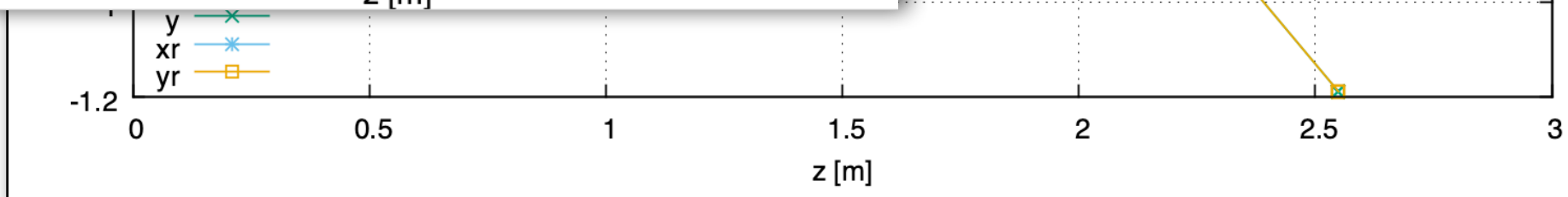
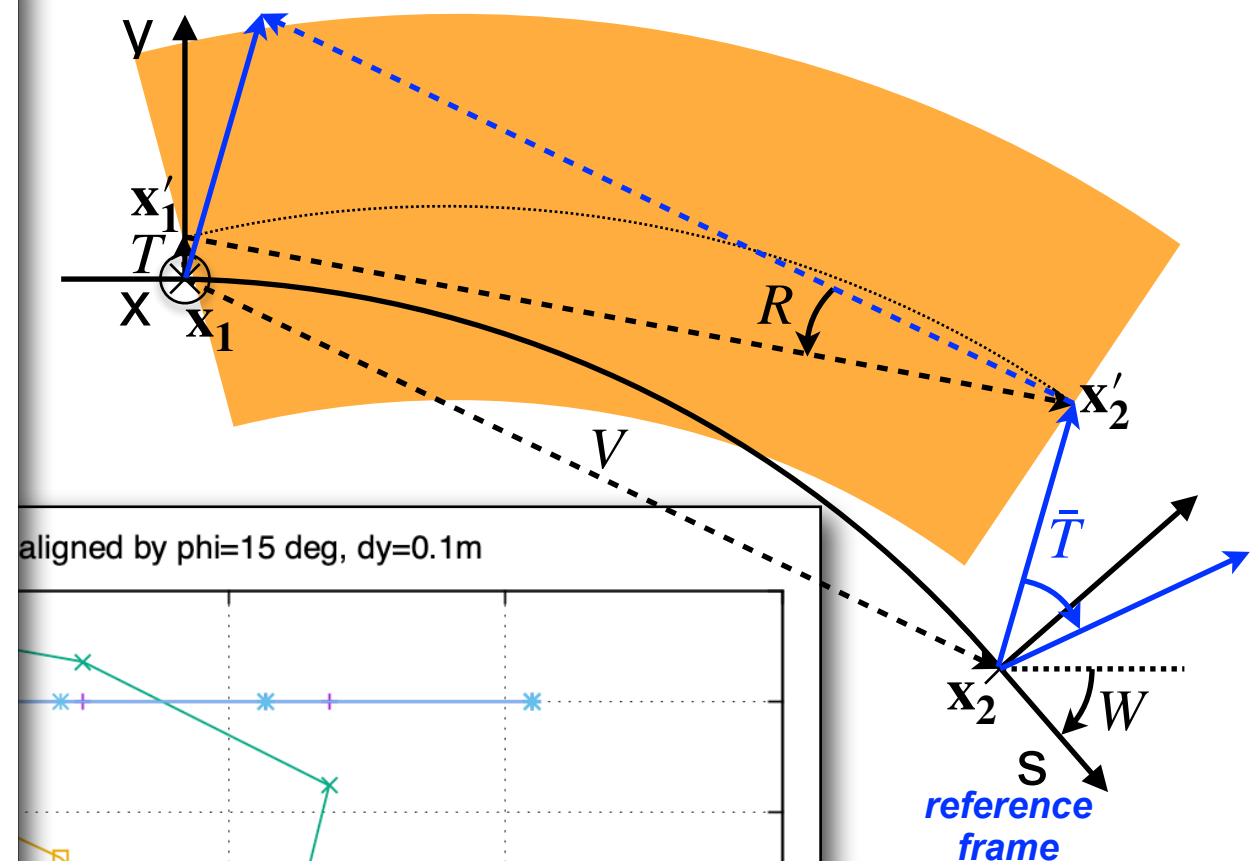
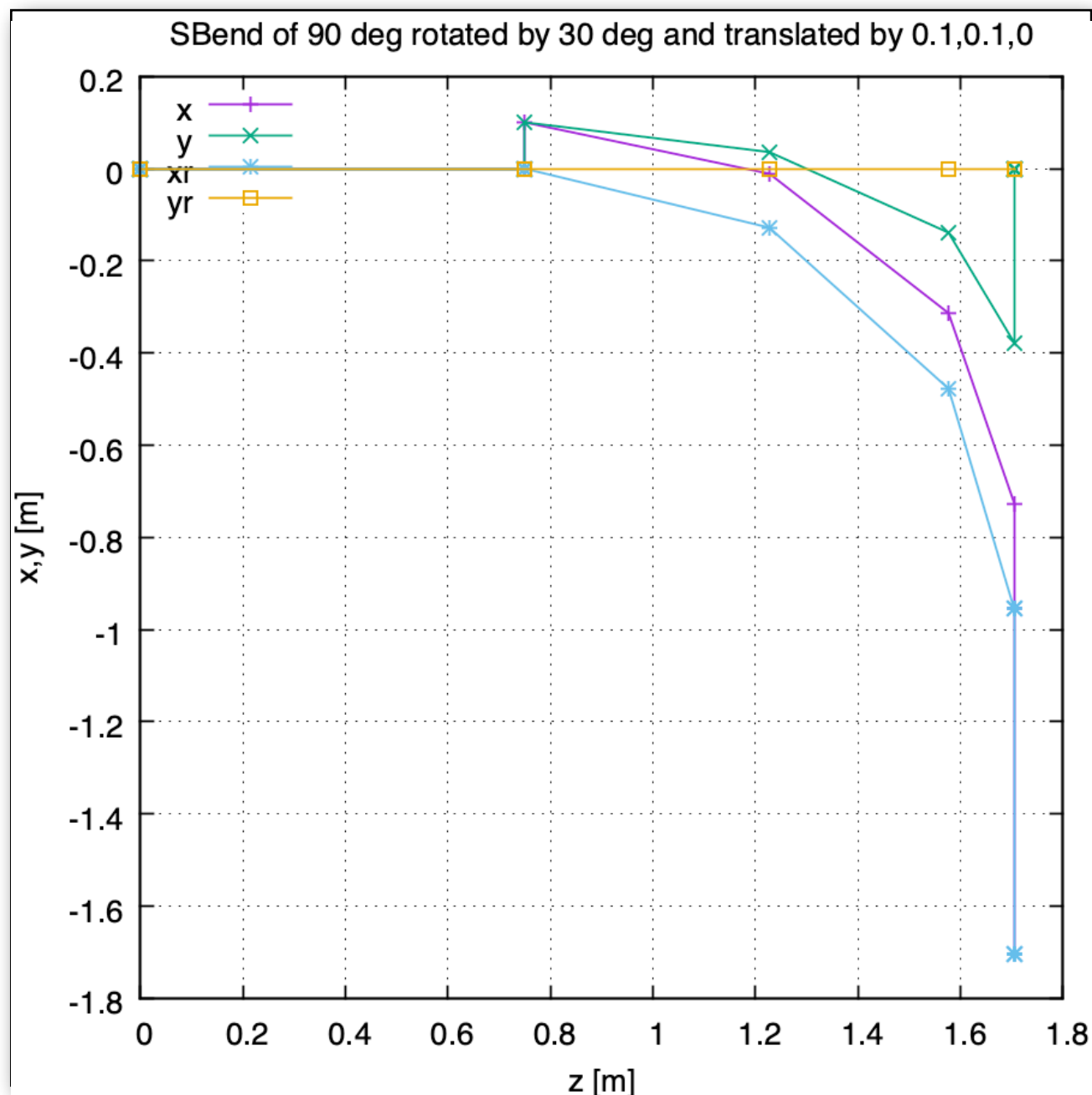
Survey: sbend tilted by 90° — $d\phi 15^\circ$ $dy 0.1m$

x, y with misalignments, xr, yr reference frame without misalignment



nslice = 3

x, y with misalignments, xr, yr reference frame without misalignment



nslice = 3



Tracking actions (*Survey, Track, Cofind and Twiss*)



- ◎ **Actions are functions** (or objects with function-like semantic).
 - ➔ MAD-NG functions are *first class lexical closures* (fun & env) and can do everything...
 - ▶ i.e. high order functions that can receive **and** return multiple arguments.
 - ➔ actions kinds: `atentry`, `atslice`, `atexit`, `ataper`, `atsave`.
 - ➔ **mechanism to customise or extend other commands** (e.g. **Twiss** with **Track** and **Cofind**).

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- ◎ Actions can be **combined** with combinators (and selectors).
 - ➔ `chain(f1,f2)` \Rightarrow `f1() ; return f2()`.
 - ➔ `achain(f1,f2)` \Rightarrow `return f1() and f2()`.
 - ➔ `ochain(f1,f2)` \Rightarrow `return f1() or f2()`.
 - ➔ `compose(f1,f2)` \Rightarrow `return f1(f2())`.
 - ➔ `ftrue, ffalse, fnone`.

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 - ➔ `ftrue, ffalse, fnone`.
- ◎ Actions can be **selected** by *selectors*:
 - ➔ Selectors are functions to enable/disable actions based on some particular criteria e.g. slices number or any other user-defined criteria.
predefined selectors: atall, atentry, atbegin, atbody, atbound, atend, atexit, atmid, atins, atstd, actionat, action.

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- ➔ `compose(f1,f2)` ➔ `return f1(f2())`.
- ➔ `ftrue`, `false`, `fnone`.

**Actions are a powerful tool to extend tracking codes (survey and track).
E.g. connect sequences (or beams) together; replace, extend or wrap computations; add extra physics between multi-particles or dampers, etc...**

- ◎ Actions can be **selected** by selectors:

- ➔ Selectors are functions to enable/disable actions e.g. slices number or any other user-defined criteria
- predefined selectors: *atall*, *atentry*, *atbegin*, *atbody*, *atbound*, *atend*, *atexit*, *atmid*, *atins*, *atstd*, *actionat*, *action*.

- ◎ **Actions are triggered by tracking codes** (**Survey** and **Track**).

- ➔ actions are chained so they are independent from each other.
- ➔ default for *ataper*: check for aperture *at slice 0 (titled frame)*.
- ➔ default for *atsave*: save data *at exit (reference frame)*,
and *at slices (titled frame)* if *atslice = ftrue*.

Order of execution at each slice

atslice = ftrue
atbegin and ataper
and ataper (user)
atsave (track)
and atsave (twiss)
and atsave (user)

Sequence

.....

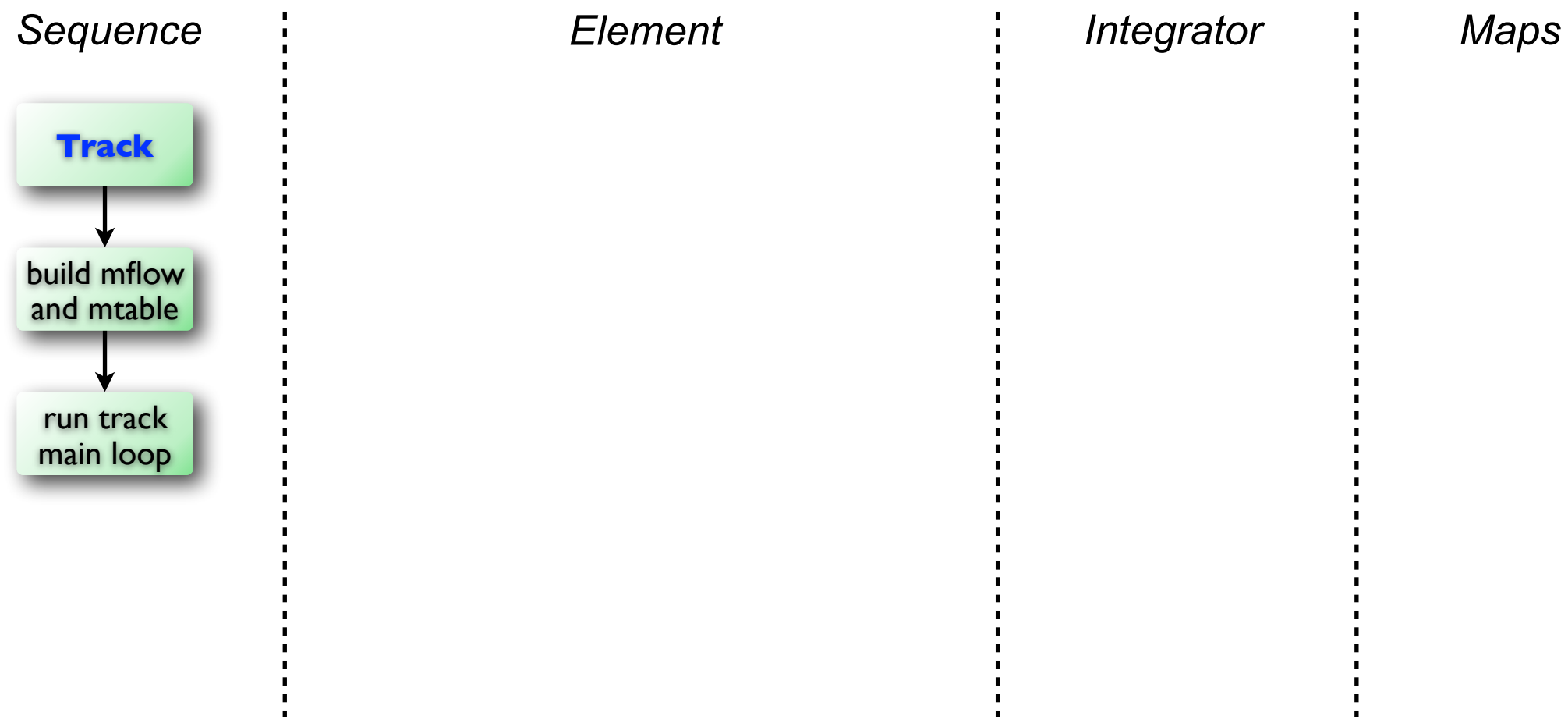
Element

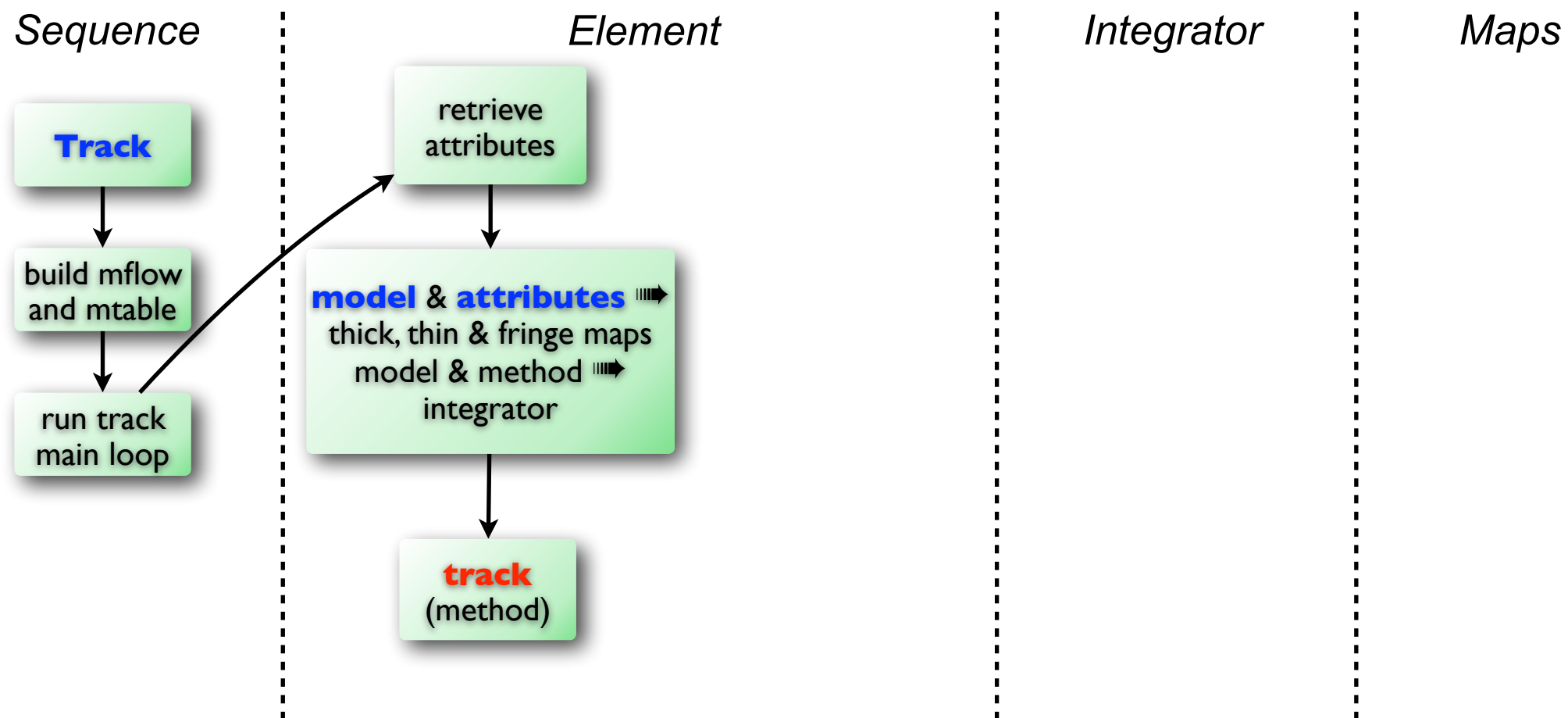
.....

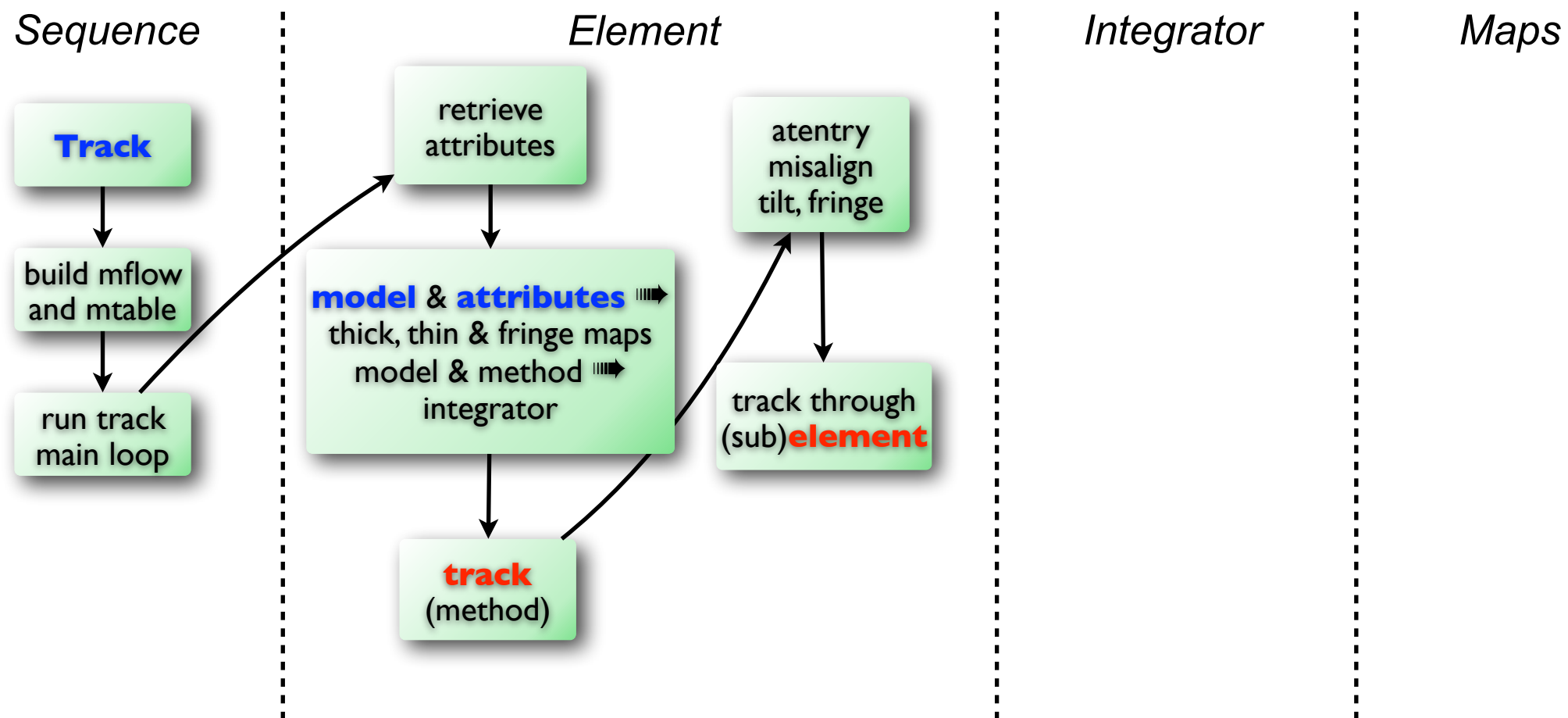
Integrator

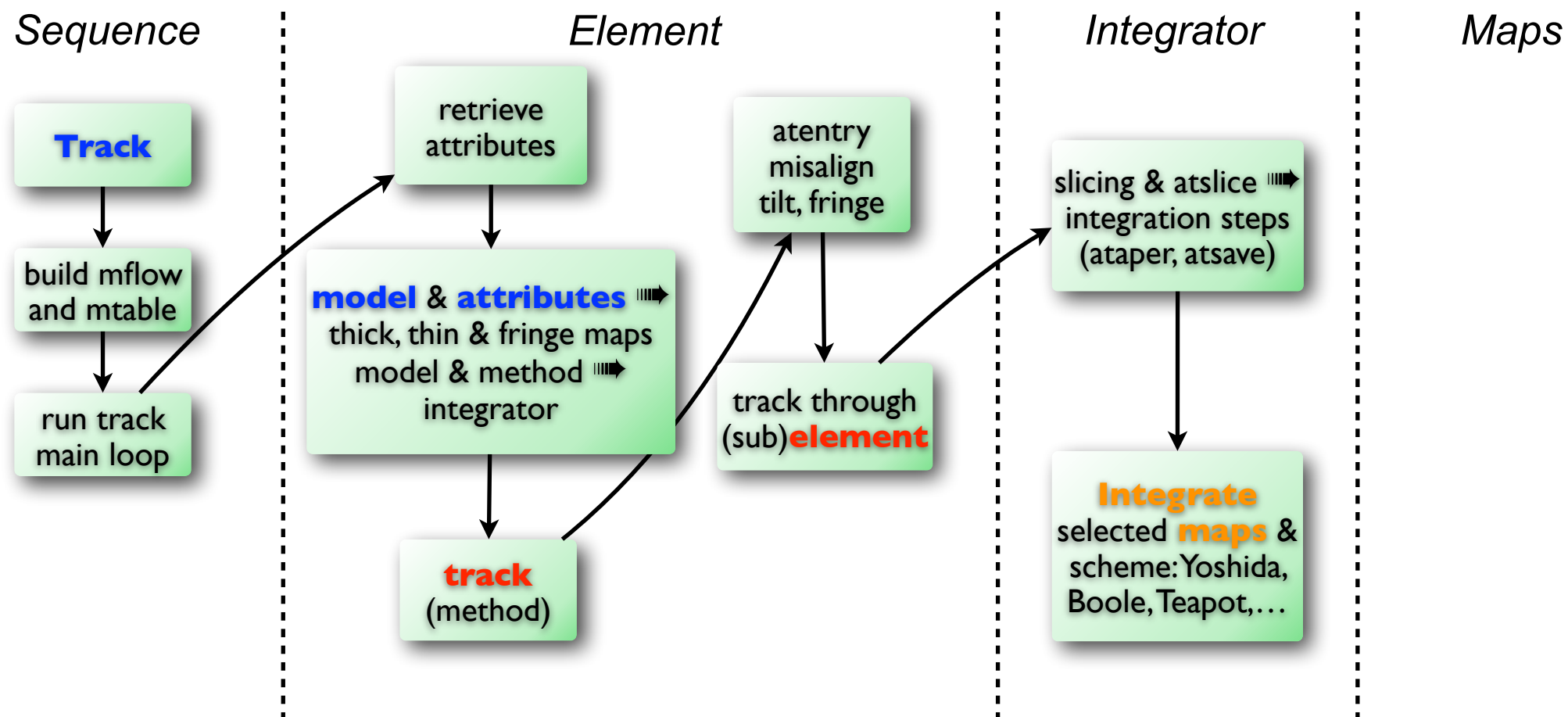
.....

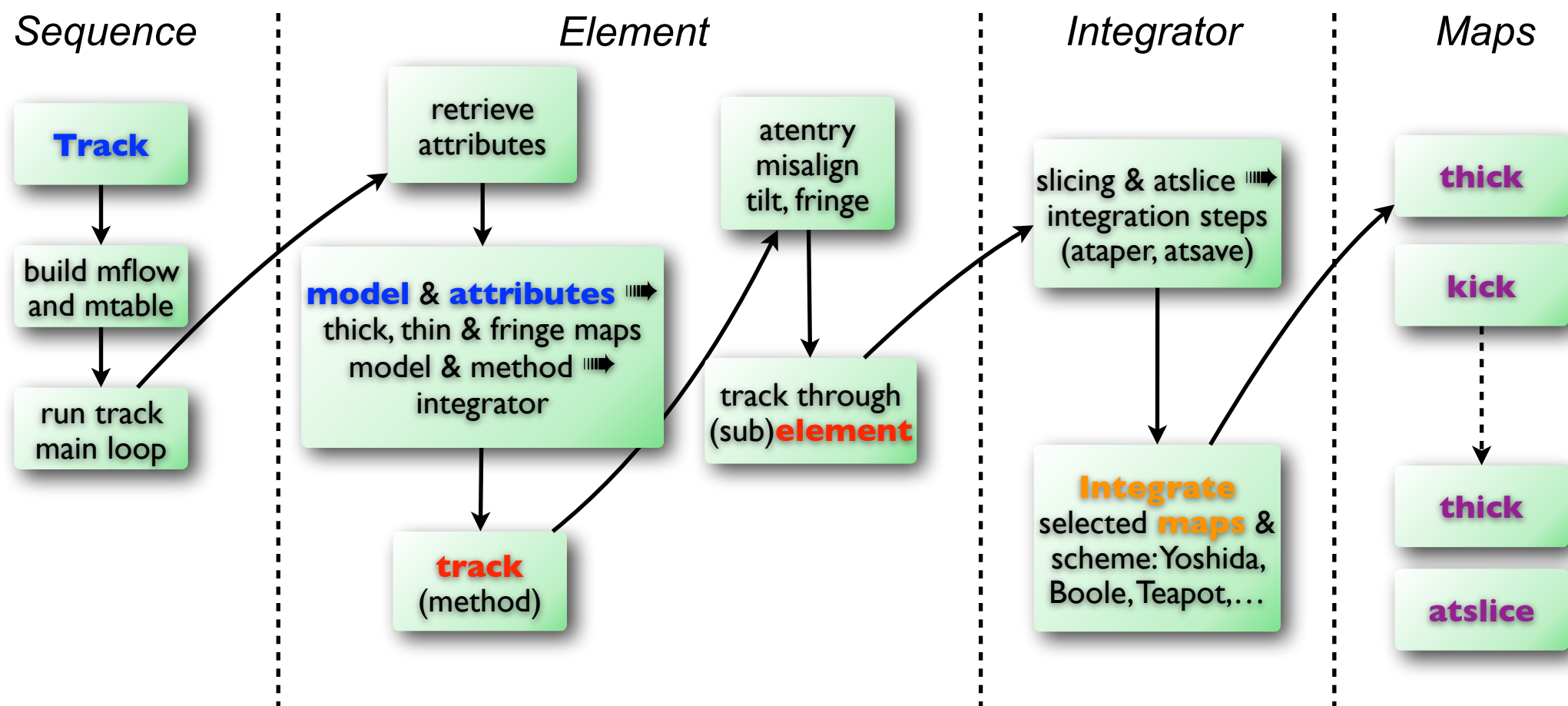
Maps

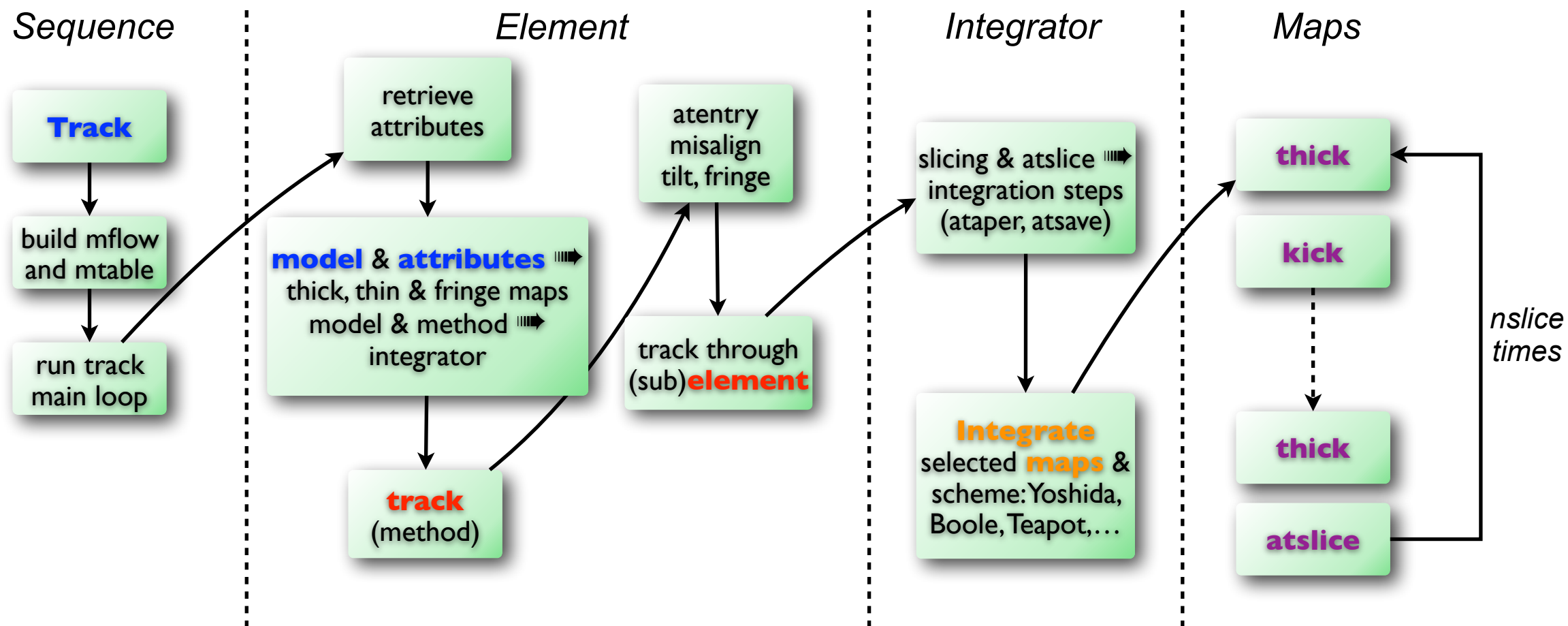


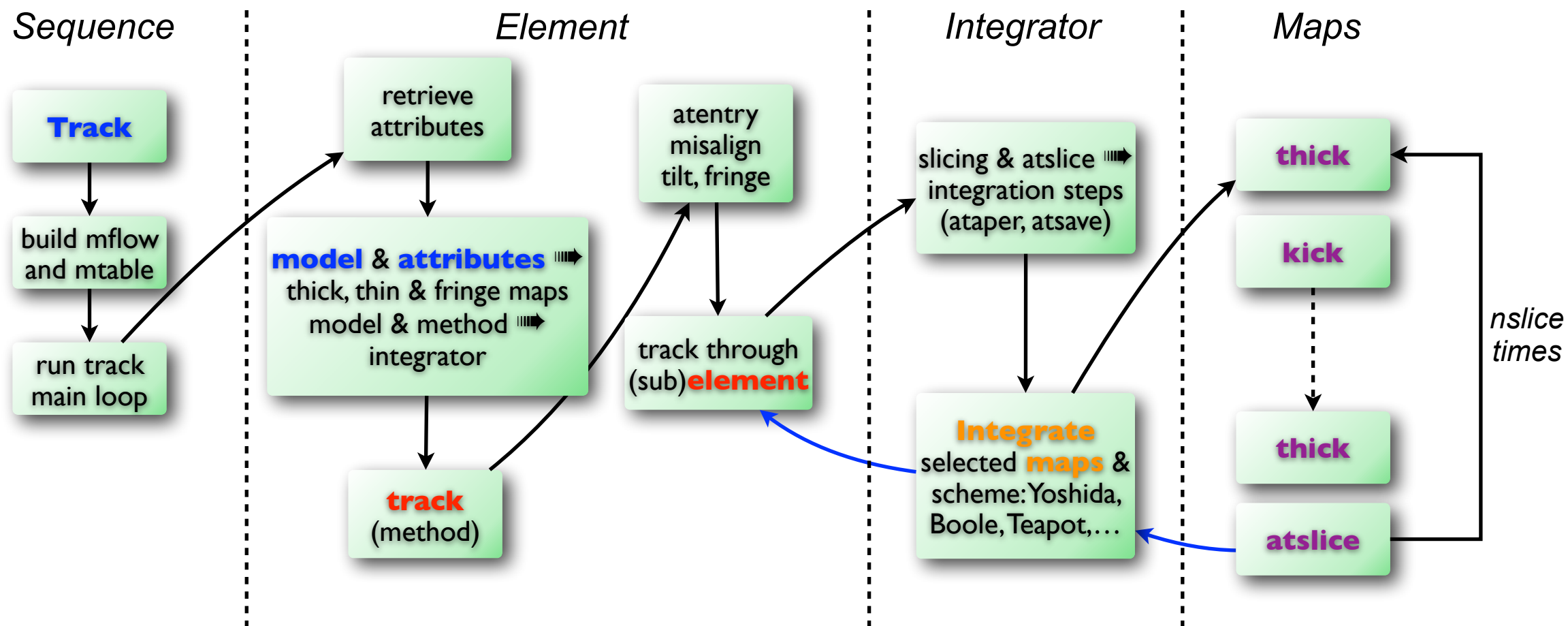


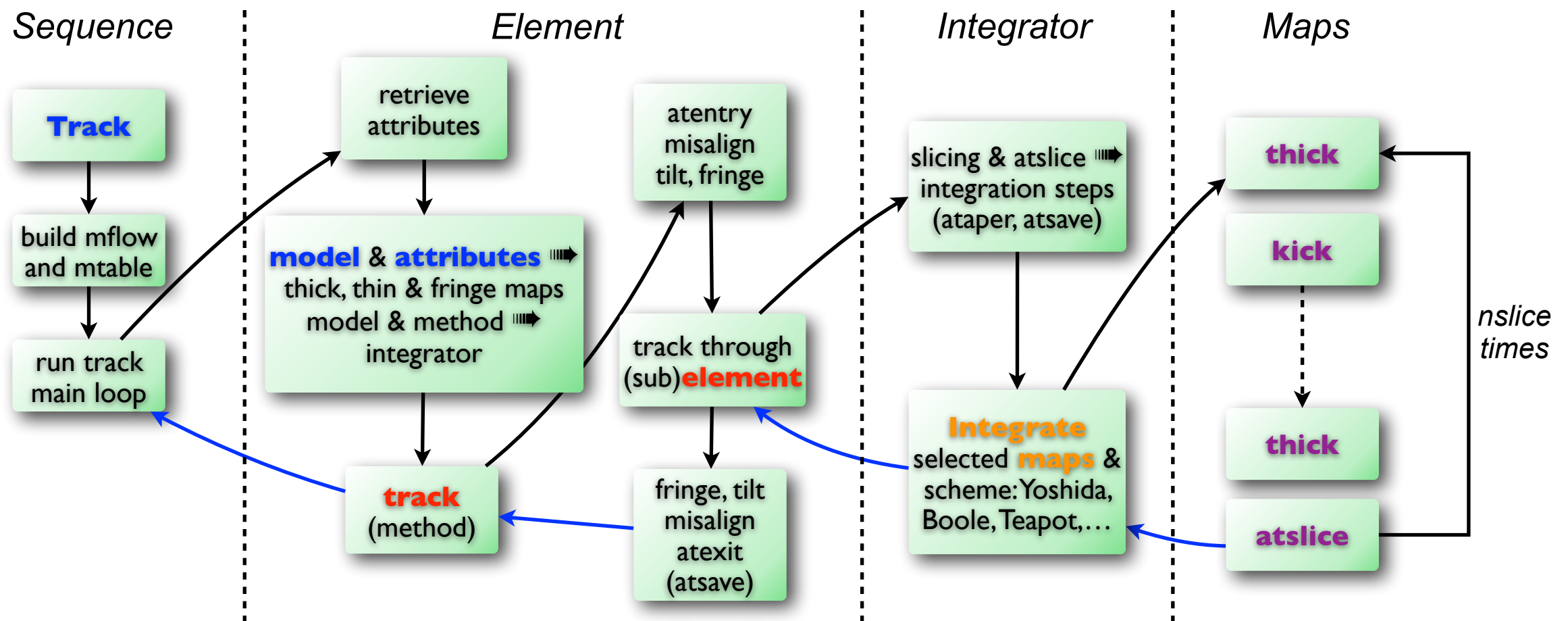


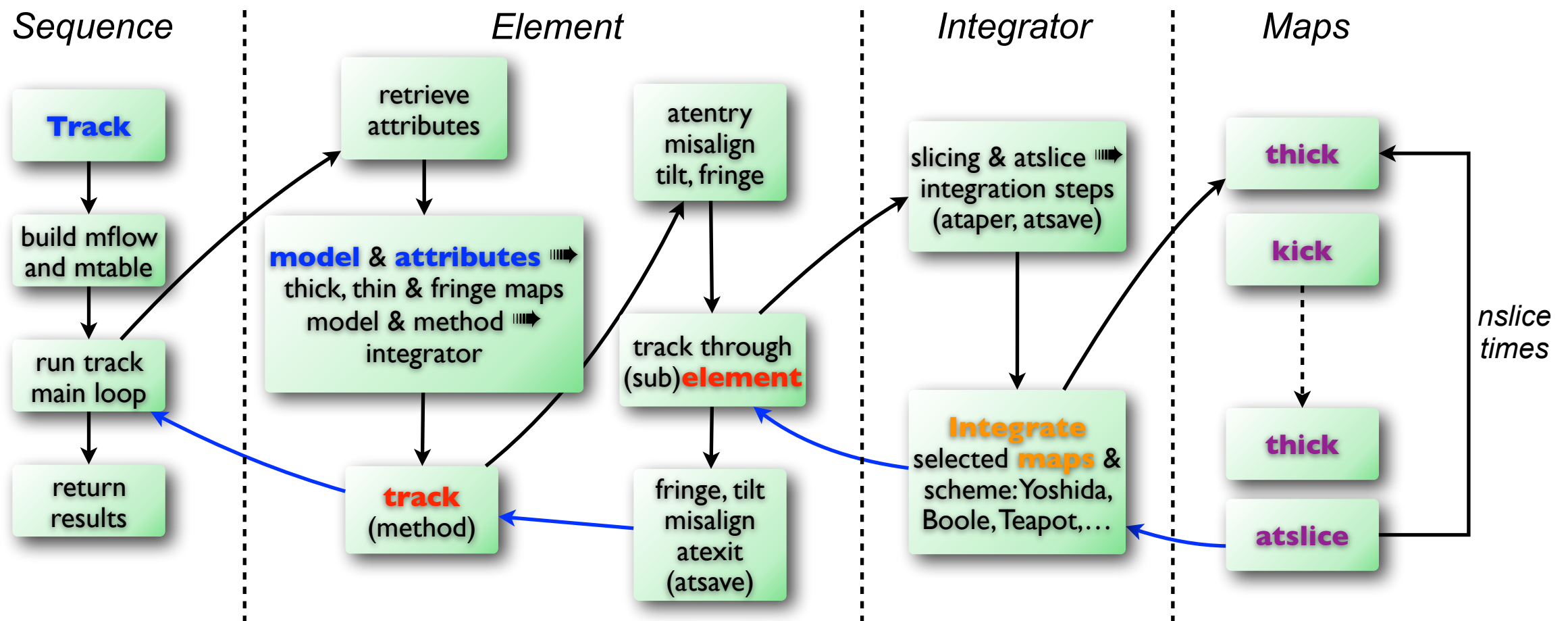


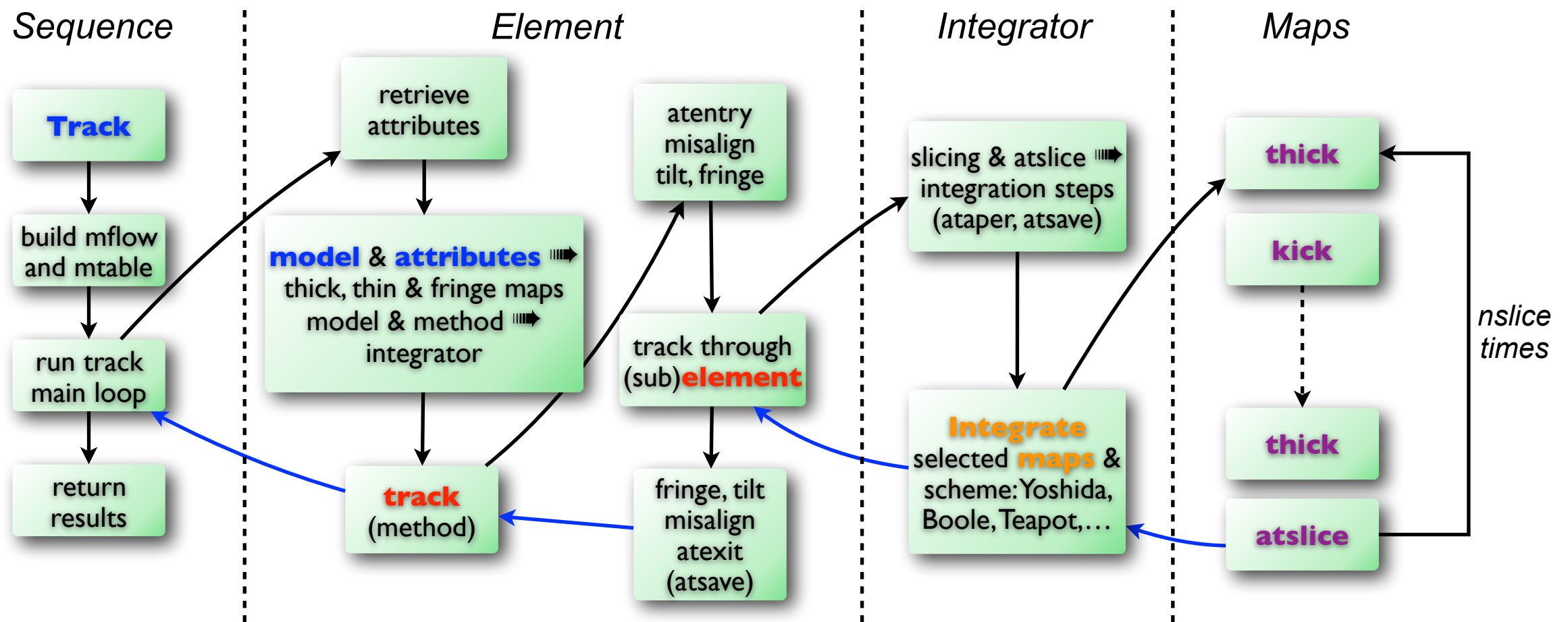




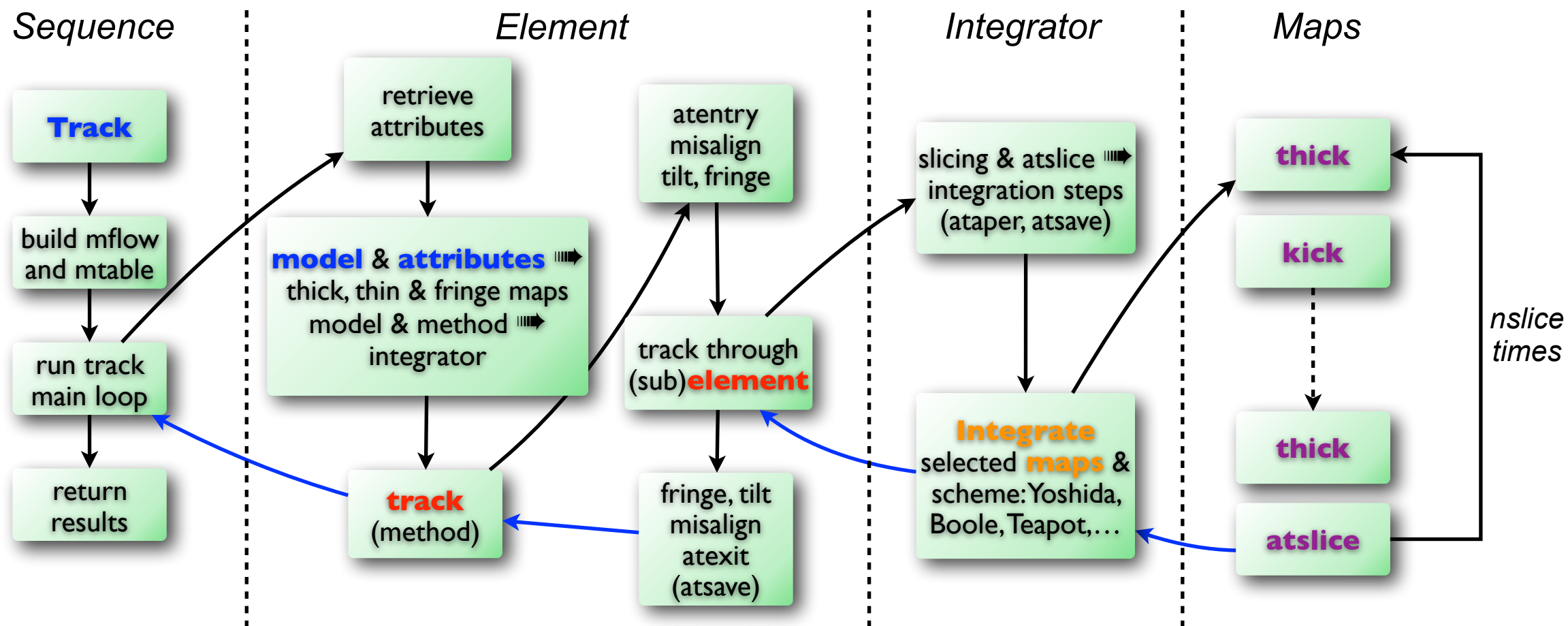






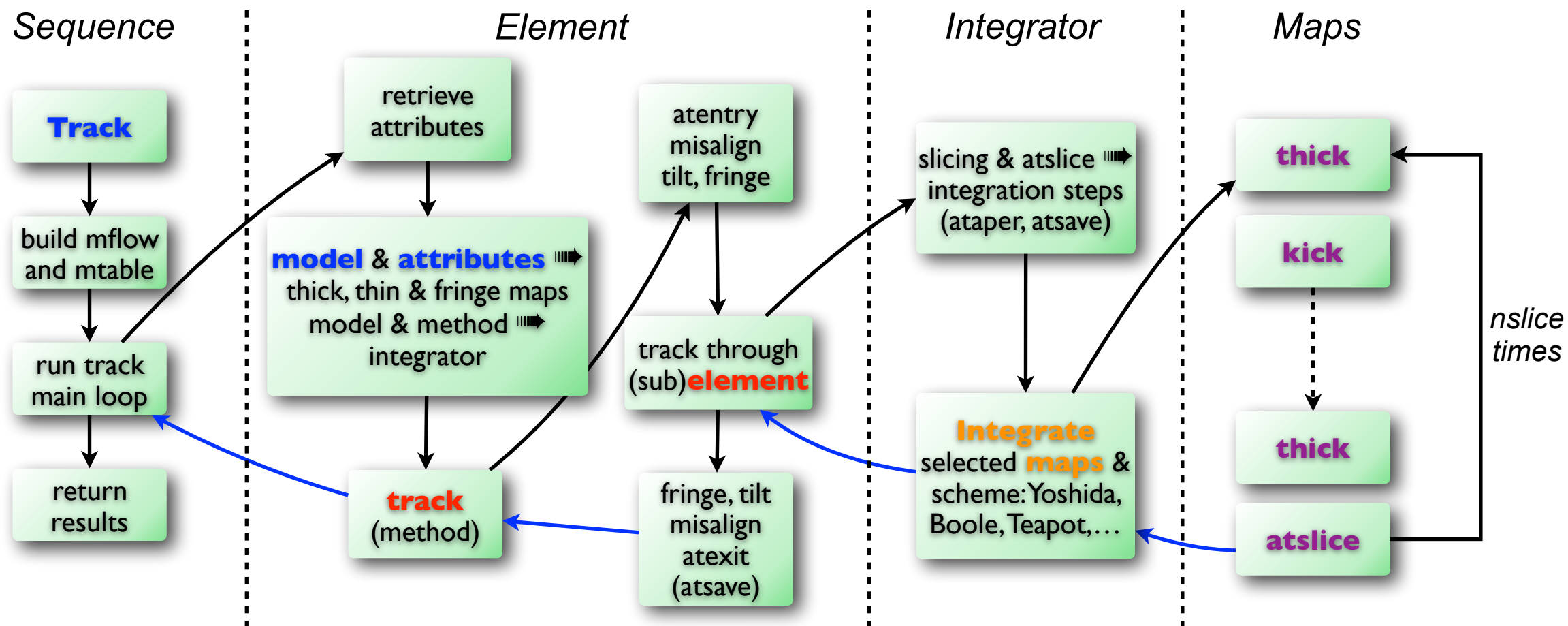


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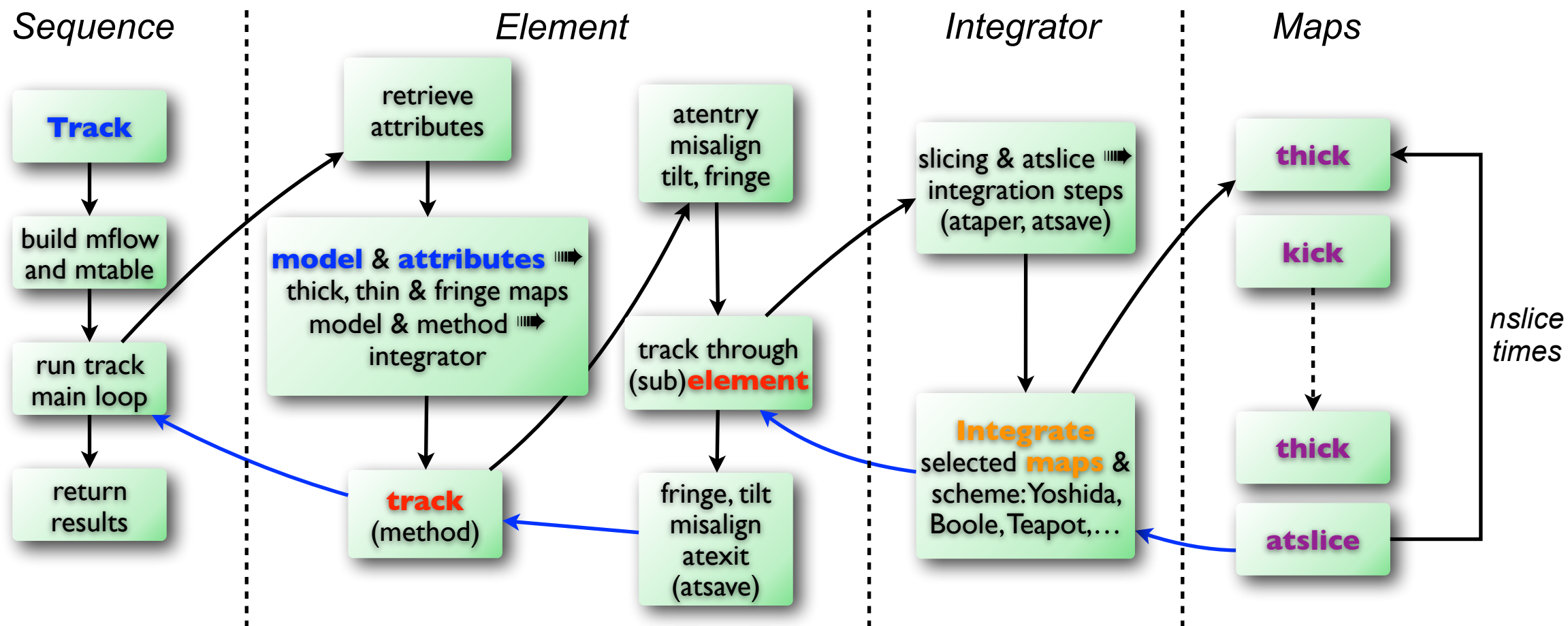
Physics can be extended by creating new element or modifying existing element or subelements track method (object oriented approach)



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Physics can be extended by providing extra integration methods e.g. 3D field maps.

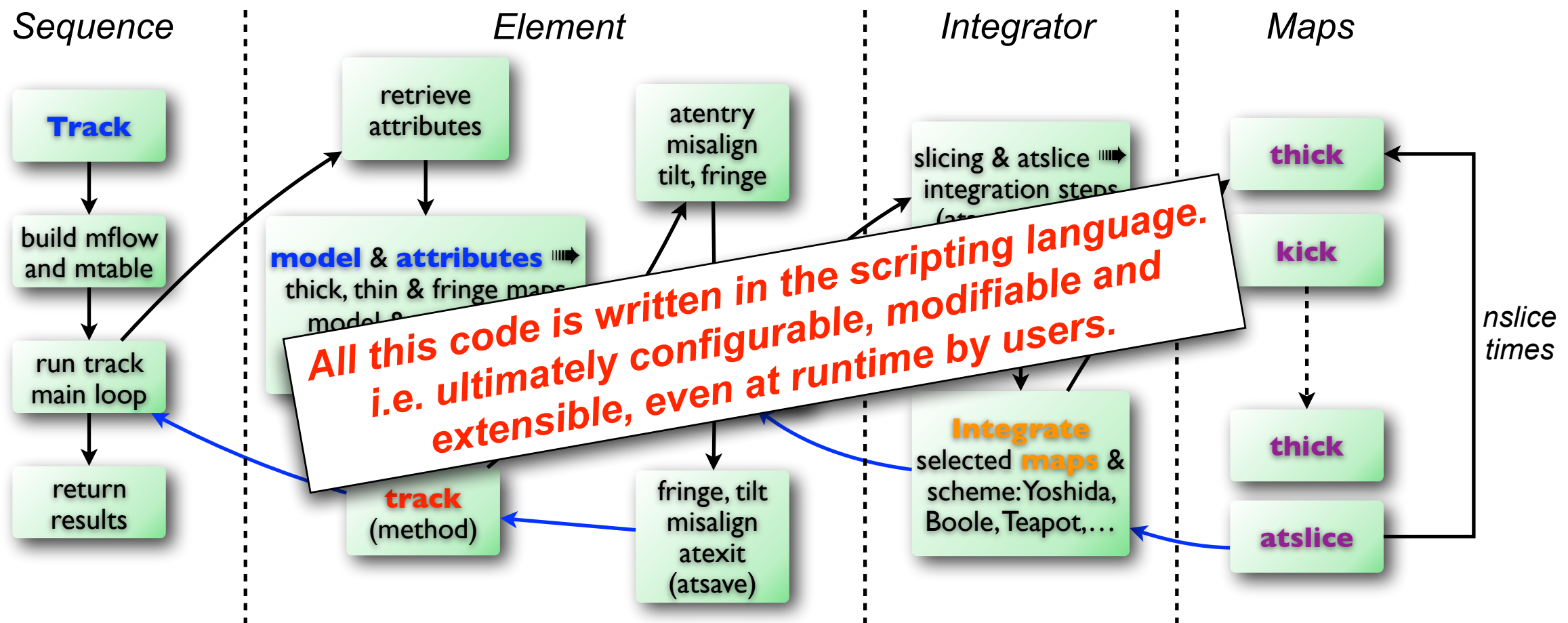


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MAD-NG physics I



- ◎ **6D PTC physics using GTPSA (for DA) and symplectic integrators.**
 - ▶ slicing, combined physics & elements, easy support for extensions, etc...
 - ▶ x4-10 faster than PTC for TPSA tracking, x1-2 slower than MAD-X for most cases.

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- ◎ **Survey: geometrical tracking**
 - ▶ Survey supports multi-turns, ranged and step-by-step **backtracking** and **reverse** tracking. Return a Survey table *and* a Survey map flow (tracked context).
 - ▶ fully compatible with Track for superposition and observable points (e.g. table output, smooth plots, slicing, actions, sub-elements, etc...)
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 - ▶ support exact misalignments and permanent misalignments, and patches.
- ◎ **Track: dynamical tracking**
 - ▶ Track supports **multi-particles** or **multi-damaps**, **multi-turns**, **ranged** and step-by-step **backtracking** and **reverse** tracking of **charged** particles to **arbitrary DA order** and arbitrary number of **parameters** (few thousands). Return a Track table *and* a Track map flow (tracked context).
 - ▶ fully compatible with Survey for superposition and observable points (same tracking engine).
 - ▶ support **exact** misalignments, **permanent** misalignments, **multipoles** & field errors **for all elements**. Can be combined freely with **patches**.
 - ▶ **symplectic tracking up to 8th order** on 5D (delta-p) and 6D (delta-rf) phase space (*exact=true, time=true, totalpath e.g. for thick RF*).
 - ▶ provides true **thick lens** and thin lens tracking model, **radiation with photons tracking** (disabled in twiss), **fringe fields** (hard edge for all elements, including solenoid), **mutable particles** (multiple beams), **exact patches** (translations, rotations & time-energy), 4D weak-strong beam-beam (sixtracklib), apertures (all kinds).
 - ▶ may search for the closed orbit to support relative initial coordinates.



MAD-NG physics II



- **Cofind:** fix point search

- ▶ Newton-based optimiser running **Track** with 1st order DA map or 7 particles.
- ▶ support final coordinates translation.
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● **Twiss**: optics tracking

- ▶ runs **Cofind** (closed orbit) - **Track** (one-turn map) - **Normal** - **Track** (optics) - post processing.
- ▶ extend Track with actions to compute on-the-fly optics and fill twiss table (extended track table).
- ▶ support coupled optics, dispersions, tunes, chromaticities, synchrotron integrals, momentum compaction factor, phase slip factor, energy gamma transition, etc... support *chrom* option to compute chromatic derivatives of previous quantities (e.g. Montague functions).

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● **Match**: highly configurable optimiser

- ▶ on the model of MAD-X use_macro approach, i.e. arbitrary user's setups & runs.
- ▶ provides all kinds of local & global, linear & non-linear, optimiser (~20 algorithms).
- ▶ very flexible, highly configurable with many physics-oriented setups (not just a penalty-function to minimise).

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 - ▶ provides few algorithms (e.g. SVD, Micado) to correct orbit using BPMs and Kickers. Supports many options.

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- **Correct: orbit correction**
 - ▶ provides few algorithms (e.g. SVD, Micado) to correct orbit using BPMs and Kickers. Supports many options.
- **Normal: normal forms analysis** (*under validation*)
 - ▶ provides linear and non-linear parametric normal forms on DA map (used by twiss) to extract RDTs. Can be applied at observable points in Track to track RDTs, either on-the-fly with actions or through post processing of DA maps saved in Track table.



MAD-NG review

- Performed from Oct. 2020 to Mar. 2021.

- ⦿ Performed from Oct. 2020 to Mar. 2021.
- ⦿ Run **simple studies** on CERN machines and compare results vs MAD-X and MADX-PTC (listed in reverse time order, from last to first).
 - ➔ *Clic 380 GeV BDS optimisation* (Andrii Pastushenko, 2 presentations)
 - twiss, high order maps generation, beam size comparison.
 - ➔ *MAD-NG outlook for LHC and HL-LHC* (Riccardo De Maria)
 - ➔ *MAD-NG in Gantries* (Cedric Hernalsteens, not presented)
 - ➔ *Experience with FCC-ee Lattice in MAD-NG* (Leon van Riesen-Haupt)
 - linear optics, momentum detuning, amplitude detuning, radiation integrals.
 - ➔ *Experience for LHC coupling with MAD-NG* (Tobias Persson).
 - example in the next slide
 - ➔ *Experience of MAD-NG with the PS* (Alexander Huschauer).
 - linear optics, dispersions, tunes, chromaticities.
 - exploration of model and integration methods.
 - ➔ *Translating MAD-X scripts to MAD-NG* (Laurent Deniau).



MAD-NG studies - LHC coupling with param. maps




```

print("strengths before matching coupling correctors:")
print("sk1r=", MADX.sk1r)
print("sk2r=", MADX.sk2r)
print("sk3r=", MADX.sk3r)
print("sk4r=", MADX.sk4r)

local X0 = damap {mo=2, nv=6, nk=4, ko=1,
                  vn={'x','px','y','py','t','pt',
                    'sk1r','sk2r','sk3r','sk4r'}}

-- set knobs: scalar + TPSA -> TPSA
MADX.sk1r = MADX.sk1r + X0.sk1r
MADX.sk2r = MADX.sk2r + X0.sk2r
MADX.sk3r = MADX.sk3r + X0.sk3r
MADX.sk4r = MADX.sk4r + X0.sk4r

local mjac = { ---> variables & knobs
  { var='x' , '0010001', '00100001', '001000001', '0010000001' }, -- |
  { var='x' , '0001001', '00010001', '000100001', '0001000001' }, -- |
  { var='px', '0010001', '00100001', '001000001', '0010000001' }, -- v
  { var='px', '0001001', '00010001', '000100001', '0001000001' }, -- constraints
}

status, fmin, ncall = match {
  command := track {sequence=lhcb1, X0=X0, observe=1, savemap=true},
  ...

```

```
print("strengths before matching coupling correctors:")
print("sk1r=", MADX.sk1r)
print("sk2r=", MADX.sk2r)
print("sk3r=", MADX.sk3r)
print("sk4r=", MADX.sk4r)

local X0 = damap {mo=2, nv=6, nk=4, ko=1,
  vn={'x','px','y','py','t','pt',
    'sk1r','sk2r','sk3r','sk4r'}}

-- set knobs: scalar + TPSA -> TPSA
MADX.sk1r = MADX.sk1r + X0.sk1r
MADX.sk2r = MADX.sk2r + X0.sk2r
MADX.sk3r = MADX.sk3r + X0.sk3r
MADX.sk4r = MADX.sk4r + X0.sk4r

local mjac = { ---> variables & knobs
  { var='x' , '0010001', '00100001', '001000001', '001000'
  { var='x' , '0001001', '00010001', '000100001', '000100'
  { var='px', '0010001', '00100001', '001000001', '001000'
  { var='px', '0001001', '00010001', '000100001', '000100'
}

status, fmin, ncall = match {
  command := track {sequence=lhcb1, X0=X0, observe=1,
  ...
```

```
status, fmin, ncall = match {
  command := track {sequence=lhcb1, X0=X0, observe=1, savemap=true},

  jacobian = \t,grd,jac => -- gradient not used, fill only jacobian
    jac:setrow(1.. 8, t['S.DS.L2.B1'].__map:getm(mjac) )
    jac:setrow(9..16, t['E.DS.L2.B1'].__map:getm(mjac) )
  end,

  variables = { rtol=1e-6, -- 1 ppm
    { name='sk1r', get := MADX.sk1r:get0(), set = \x -> MADX.sk1r:set0(x) },
    { name='sk2r', get := MADX.sk2r:get0(), set = \x -> MADX.sk2r:set0(x) },
    { name='sk3r', get := MADX.sk3r:get0(), set = \x -> MADX.sk3r:set0(x) },
    { name='sk4r', get := MADX.sk4r:get0(), set = \x -> MADX.sk4r:set0(x) },
  },

  equalities = {
    { expr = \t -> t['S.DS.L2.B1'].__map.x :get'0010', name='S.R11.x' },
    { expr = \t -> t['S.DS.L2.B1'].__map.x :get'0001', name='S.R12.x' },
    { expr = \t -> t['S.DS.L2.B1'].__map.px:get'0010', name='S.R21.x' },
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  },

  objective = { fmin=1e-12 },
  maxcall=100, info=2
}

-- reset knobs: extract scalar values from TPSA
MADX.sk1r = MADX.sk1r:get0()
MADX.sk2r = MADX.sk2r:get0()
MADX.sk3r = MADX.sk3r:get0()
MADX.sk4r = MADX.sk4r:get0()

print("status=", status, "fmin=", fmin, "ncall=", ncall)
print("strengths after matching coupling correctors:")
print("sk1r=" , MADX.sk1r )
print("sk2r=" , MADX.sk2r )
print("sk3r=" , MADX.sk3r )
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```

```
print("strengths before matching coupling correctors:")
print("sk1r=", MADX.sk1r)
print("sk2r=", MADX.sk2r)
print("sk3r=", MADX.sk3r)
print("sk4r=", MADX.sk4r)

local X0 = damap {mo=2, nv=6, nk=4, ko=1,
    vn={'x','px','y','py','t','pt',
        'sk1r','sk2r','sk3r','sk4r'}}

-- set knobs: scalar + TPSA -> TPSA
MADX.sk1r = MADX.sk1r + X0.sk1r
MADX.sk2r = MADX.sk2r + X0.sk2r
MADX.sk3r = MADX.sk3r + X0.sk3r
MADX.sk4r = MADX.sk4r + X0.sk4r

local mjac = { ----> variables & knobs
    { var='x' , '0010001', '00100001', '001000001', '001000'
    { var='x' , '0001001', '00010001', '000100001', '000100'
    { var='px', '0010001', '00100001', '001000001', '001000'
    { var='px', '0001001', '00010001', '000100001', '000100'
    }

status, fmin, ncall = match {
    command := track {sequence=lhcb1, X0=X0, observe=1,
    ...
```

```
status, fmin, ncall = match {
    command := track {sequence=lhcb1, X0=X0, observe=1, savemap=true},

    jacobian = \t,grd,jac => -- gradient not used, fill only jacobian
        jac:setrow(1.. 8, t['S.DS.L2.B1'].__map:getm(mjac) )
        jac:setrow(9..16, t['E.DS.L2.B1'].__map:getm(mjac) )
    end,

    variables = { rtol=1e-6, -- 1 ppm
        { name='sk1r', get := MADX.sk1r:get0(), set = \x -> MADX.sk1r:set0(x) },
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```

Timing summary and links to codes:

MAD-X using matrix 1m55

MAD-NG using matrix 55s (15s)

MAD-NG using matrix & knobs 40s (4.5s)

MADX-PTC using alphas-betas >40m

```
print("strengths before matching coupling correctors:")
print("sk1r=", MADX.sk1r)
print("sk2r=", MADX.sk2r)
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local X0 = damap {mo=2, nv=6, nk=4, ko=1,
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MADX.sk1r = MADX.sk1r + X0.sk1r
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MAD-NG using matrix 55s (15s)

MAD-NG using matrix & knobs 4s

MADX-PTC using

Match command performs a Principal Component Analysis on the Jacobian and tags useless constraints and variables, i.e. starting with oversized set of knobs or constrains does not harm when using parametric maps!

reset knobs: extract scalar values from TPSA

```
.sk1r = MADX.sk1r:get0()
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Conclusions

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- 2022 will focus on participation to real studies and consolidation.
 - ▶ bottom-top validation for the physics of real case studies.
 - ▶ add missing physics on demand (e.g. tapering, spin, generalised multipoles).
 - ▶ complete unit tests & manual.
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- On some aspects, MAD-NG is more mature than MAD-X
 - ▶ better code architecture and structure.
 - ▶ more flexible and extensible for the physics (new features require day(s)).
 - ▶ less surprises when combining features (e.g. misalignments and slicing).
 - ▶ main stream programming language for scripting (save user time!) & many toolboxes.
 - ▶ mature technologies, syntax error, backtrace, debugger, profiler, JIT (save user time!).
 - ▶ some features have been back ported to MAD-X (e.g. permanent misalignment, patches) or will be (fringe fields, combined/overlapping elements).
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Do not hesitate to ask me some help!

THANK YOU FOR YOUR ATTENTION

MAD scripting language is based on Lua 5.1+ (it is a superset of)



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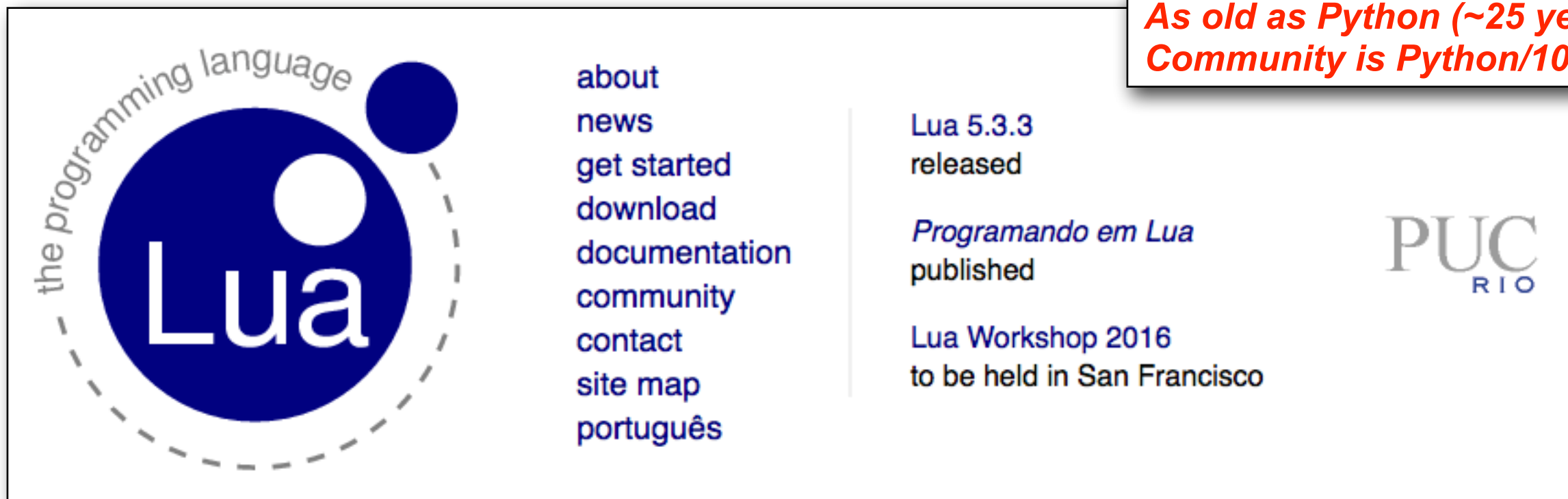
Lua is a powerful, fast, lightweight, embeddable scripting language.

Lua combines simple procedural syntax with powerful data description constructs based on **associative arrays** and **extensible semantics**. Lua is **dynamically typed**, runs by interpreting bytecode for a register-based virtual machine, and has automatic memory management with incremental garbage collection, making it **ideal for configuration, scripting, and rapid prototyping**.

Lua has been used in many industrial applications (e.g., Adobe's Photoshop Lightroom), with an emphasis on embedded systems (e.g., the Ginga middleware for digital TV in Brazil) and games (e.g., World of Warcraft and Angry Birds). Lua is currently the leading scripting language in games. **Lua has a solid reference manual and there are several books about it.** Several versions of Lua have been released and used in real applications since its **creation in 1993**. Lua featured in HOPL III, the Third ACM SIGPLAN History of Programming Languages Conference, in June 2007. Lua won the Front Line Award 2011 from the Game Developers Magazine.

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The screenshot shows the Lua website layout. On the left is the Lua logo, a blue circle with a white 'L' and 'ua' inside, and the text 'the programming language' curved around it. To the right of the logo is a vertical list of links: 'about', 'news', 'get started', 'download', 'documentation', 'community', 'contact', 'site map', and 'português'. Further right, there are three announcements: 'Lua 5.3.3 released', 'Programando em Lua published' (with the PUC RIO logo to its right), and 'Lua Workshop 2016 to be held in San Francisco'.

Reference manual is 29 pages!

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LuaJIT is Copyright © 2005-2015 Mike Pall, released under the [MIT open source license](#).

Compatibility

Windows	Linux	BSD	OSX	POSIX	
Embedded	Android	IOS			
PS3	PS4	PS Vita	Xbox 360		
GCC	CLANG LLVM	MSVC			
x86	x64	ARM	PPC	e500	MIPS
Lua 5.1 API+ABI	+ JIT	+ BitOp	+ FFI	Drop-in DLL/.so	

Overview

3x - 100x	115 KB VM	90 KB JIT	63 KLOC C	24 KLOC ASM	11 KLOC Lua
-----------	-----------	-----------	-----------	-------------	-------------

LuaJIT has been successfully used as a **scripting middleware** in games, appliances, network and graphics apps, numerical simulations, trading platforms and many other specialty applications. It scales from embedded devices, smartphones, desktops up to server farms. It combines high flexibility with **high performance** and an unmatched **low memory footprint**.

LuaJIT has been in continuous development since 2005. It's widely considered to be **one of the fastest dynamic language implementations**. It has outperformed other dynamic languages on many cross-language benchmarks since its first release — often by a substantial margin. For **LuaJIT 2.0**, the whole VM has been rewritten from the ground up and relentlessly optimised for performance. It combines a **high-speed interpreter**, written in assembler, with a **state-of-the-art JIT compiler**.

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GTPSA in a nutshell

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1 variable x at order n in the *neighbourhood* of the point a in the domain of the function f :

$$T_f^n(x; a) = f(a) + f'(a)(x - a) + \frac{f''(a)}{2!}(x - a)^2 + \dots + \frac{f^{(n)}(a)}{n!}(x - a)^n = \sum_{k=0}^n \frac{f_a^{(k)}}{k!}(x - a)^k$$

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TPSA coefficients

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convergence of the remainder (i.e. truncation error):

$$\lim_{n \rightarrow \infty} R_f^n(x; a) = \lim_{n \rightarrow \infty} f(x) - T_f^n(x; a) = 0$$

TPSA coefficients

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$f(x)$ is an analytic function, $T_f^n(x; a)$ is a polynomial approximation nearby a with radius of convergence h : $\min_{h>0} \lim_{n \rightarrow \infty} R_f^n(a \pm h; a) \neq 0$.

2 variables (x, y) at order 2 nearby (a, b) :

$$T_f^2(x, y; a, b) = f(a, b) + \frac{\partial f}{\partial x} \Big|_{(a,b)} (x - a) + \frac{\partial f}{\partial y} \Big|_{(a,b)} (y - b) + \dots$$

homogeneous polynomials

f must not depend on the integration path, i.e. must derive from a potential!

$$+ \frac{1}{2!} \left(\frac{\partial^2 f}{\partial x^2} \Big|_{(a,b)} (x - a)^2 + 2 \frac{\partial^2 f}{\partial x \partial y} \Big|_{(a,b)} (x - a)(y - b) + \frac{\partial^2 f}{\partial y^2} \Big|_{(a,b)} (y - b)^2 \right)$$

ν variables X at order n nearby A :

TPSA coefficients

$$T_f^n(X; A) = \sum_{k=0}^n \frac{f_A^{(k)}}{k!} (X; A)^k = \sum_{k=0}^n \frac{1}{k!} \sum_{|\vec{m}|=k} \binom{k}{\vec{m}} \frac{\partial^k f}{\partial X^{\vec{m}}} \Big|_A (X; A)^{\vec{m}} \quad \text{with} \quad \binom{k}{\vec{m}} = \frac{k!}{c_1! c_2! \dots c_\nu!}$$

monomials of order k

multinomial



Accuracy of TPSA (myths and legends)

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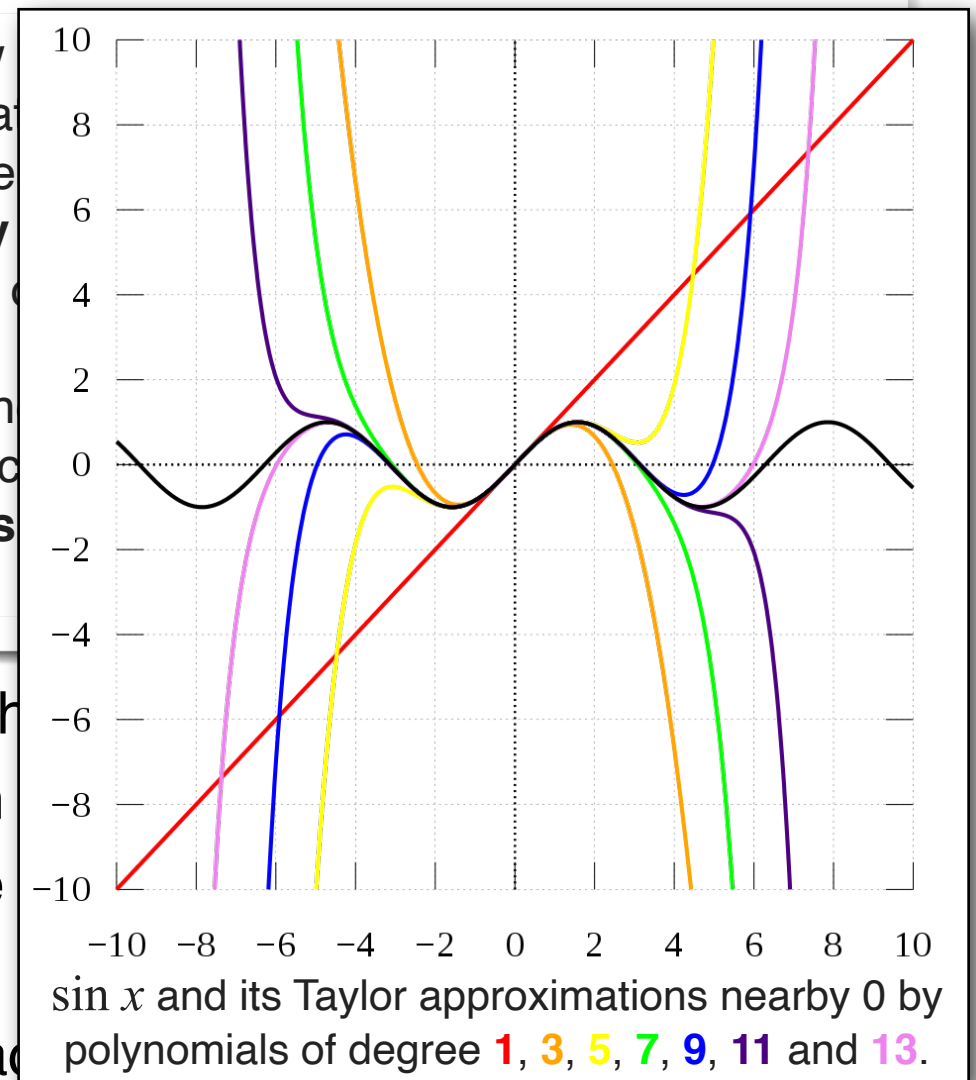
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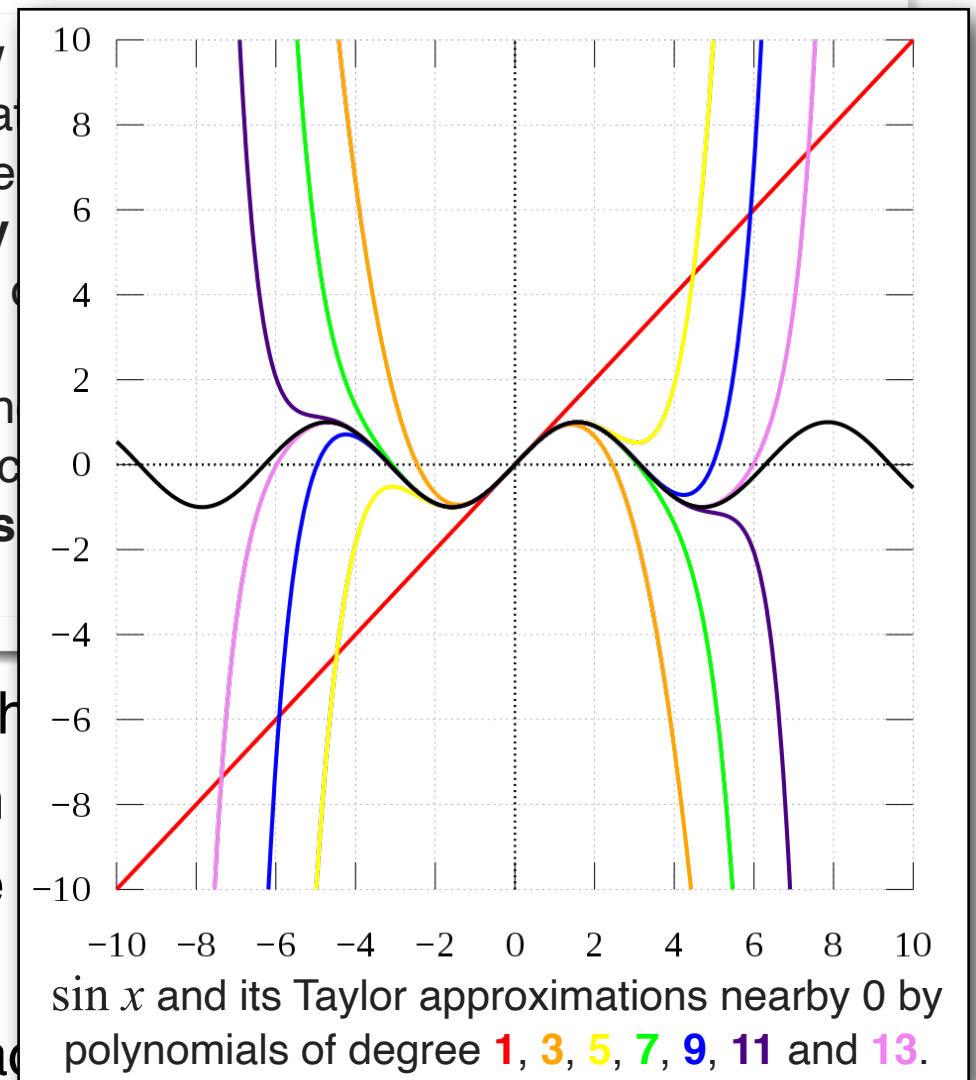
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DA map

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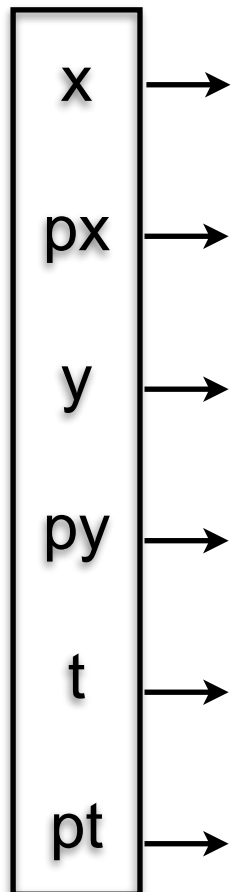
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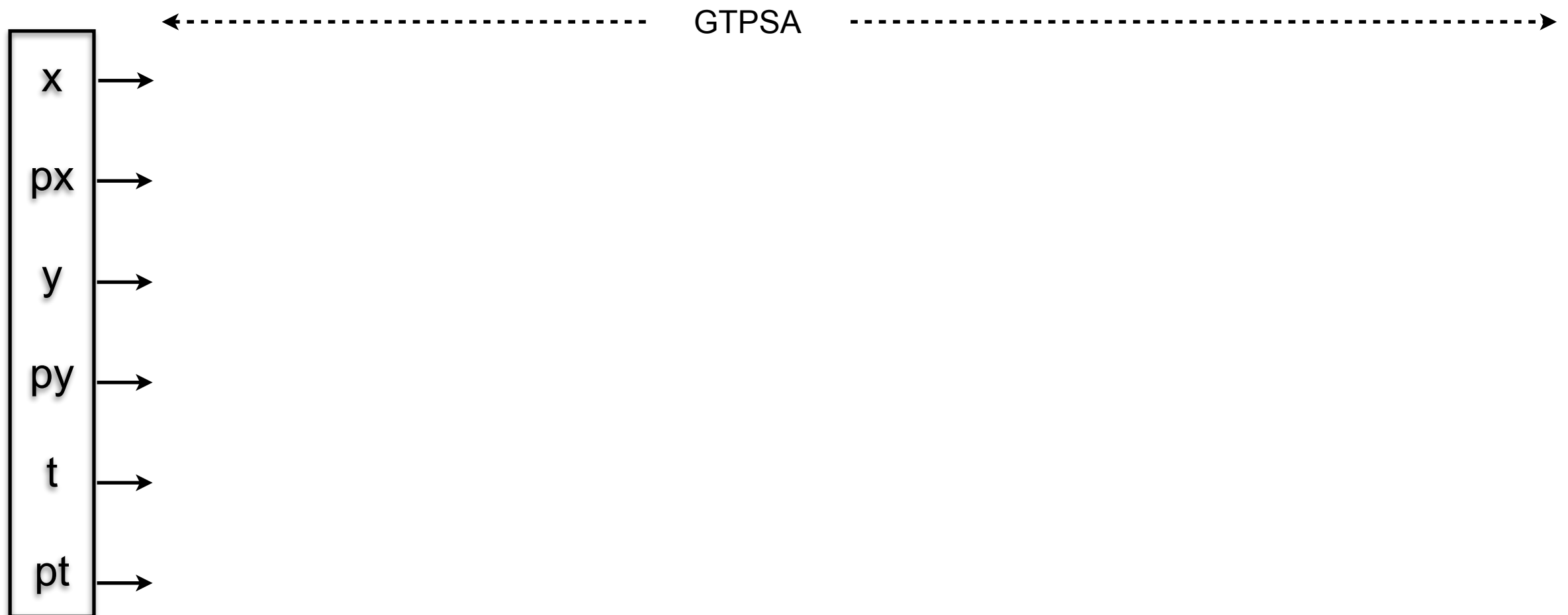
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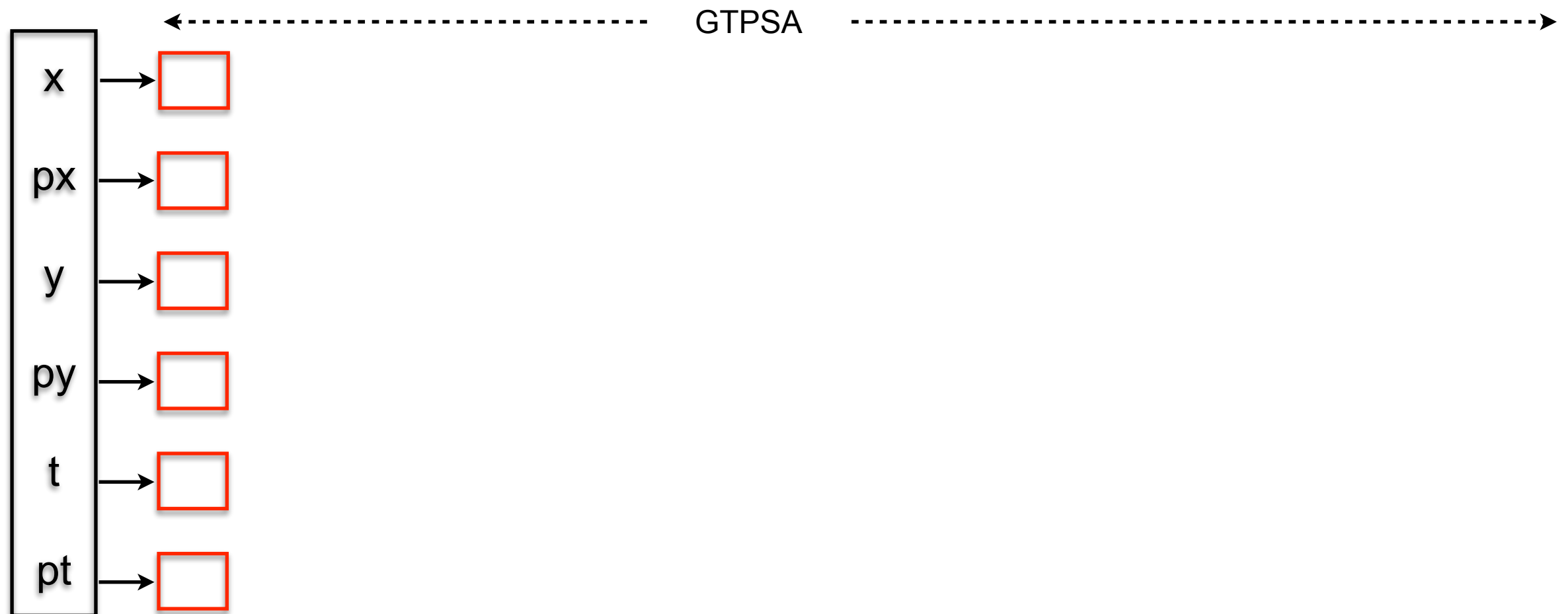
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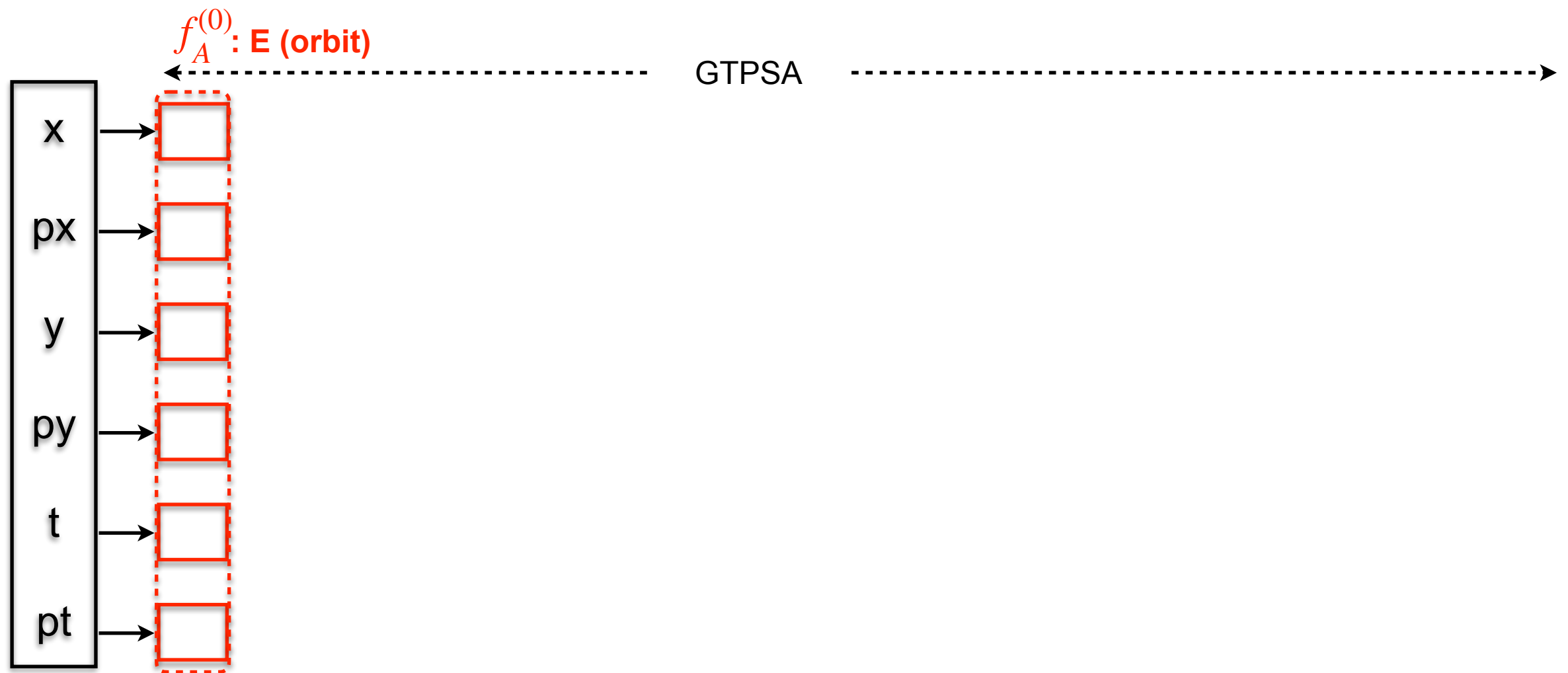
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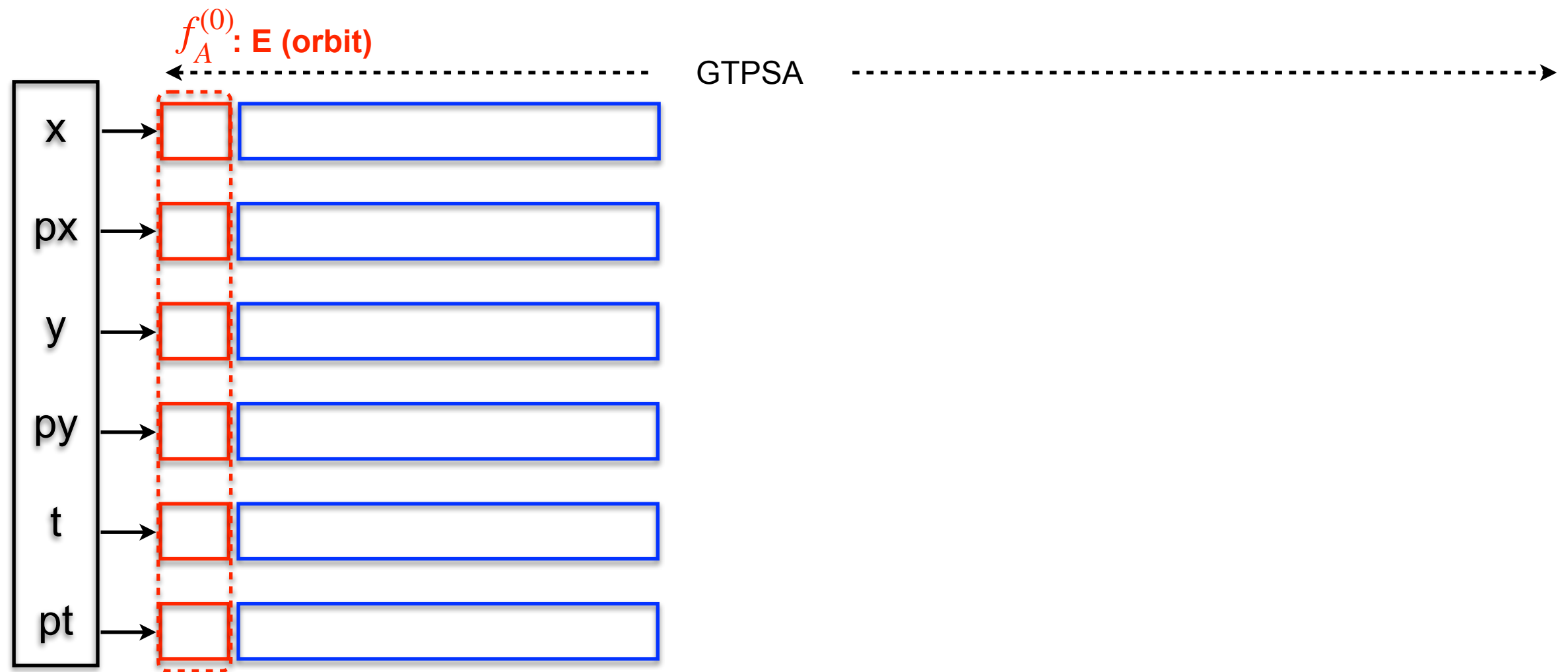
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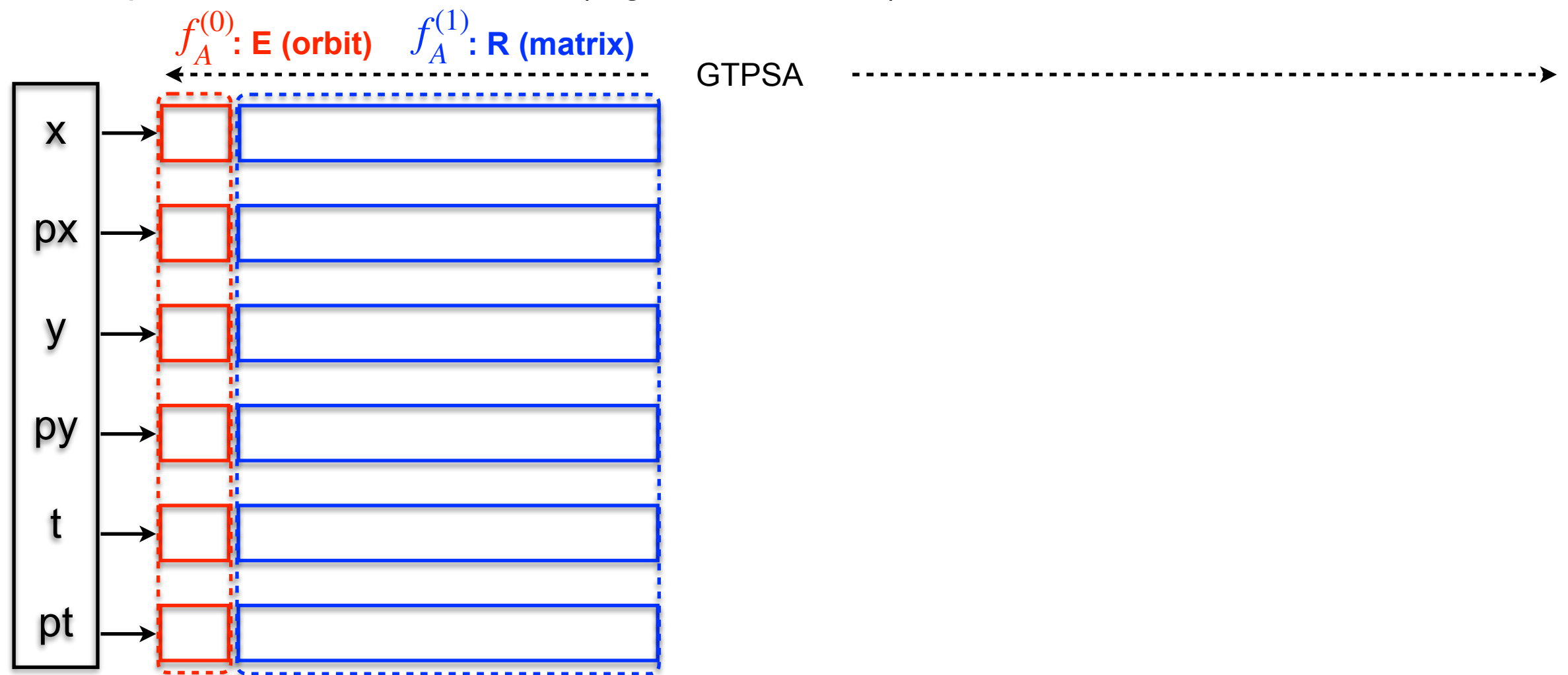
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DA map of 6 variables at order 2 (e.g. MAD-X twiss)



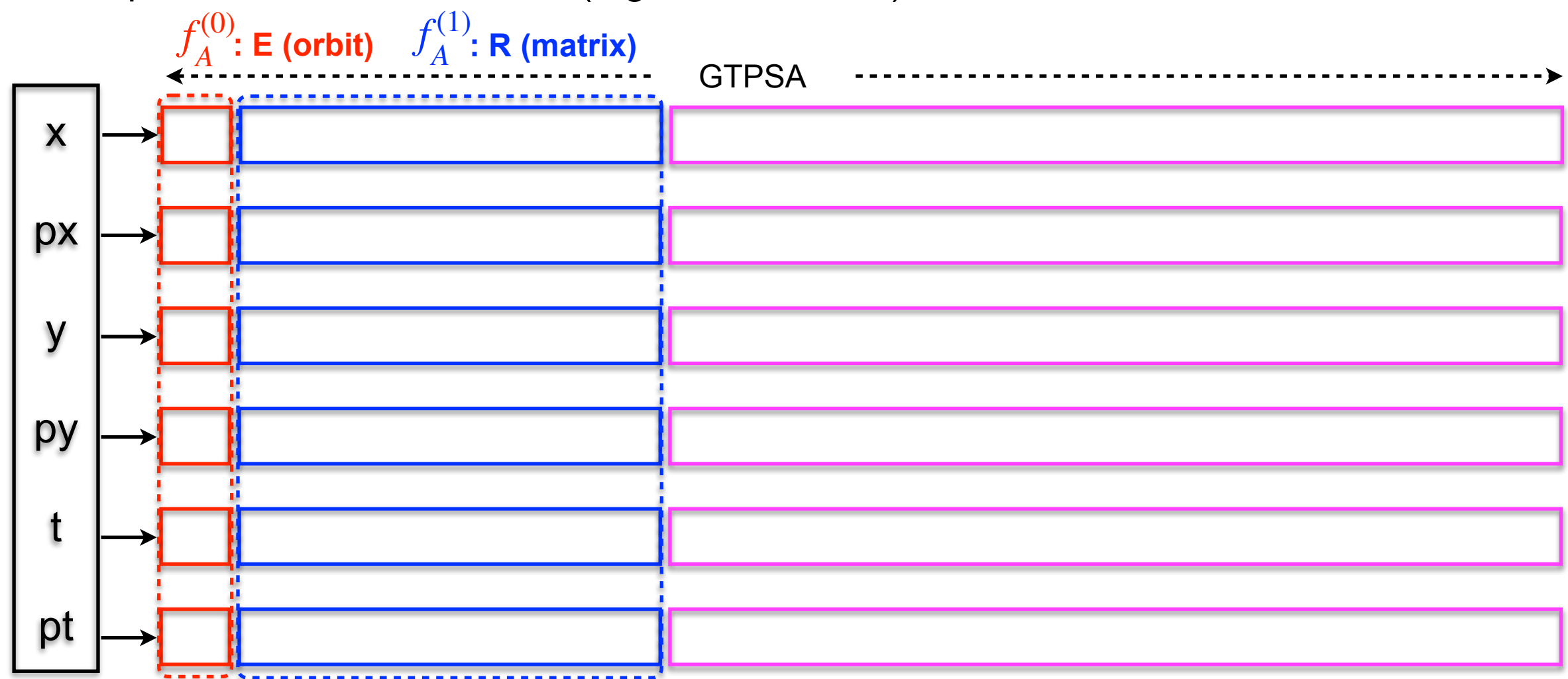
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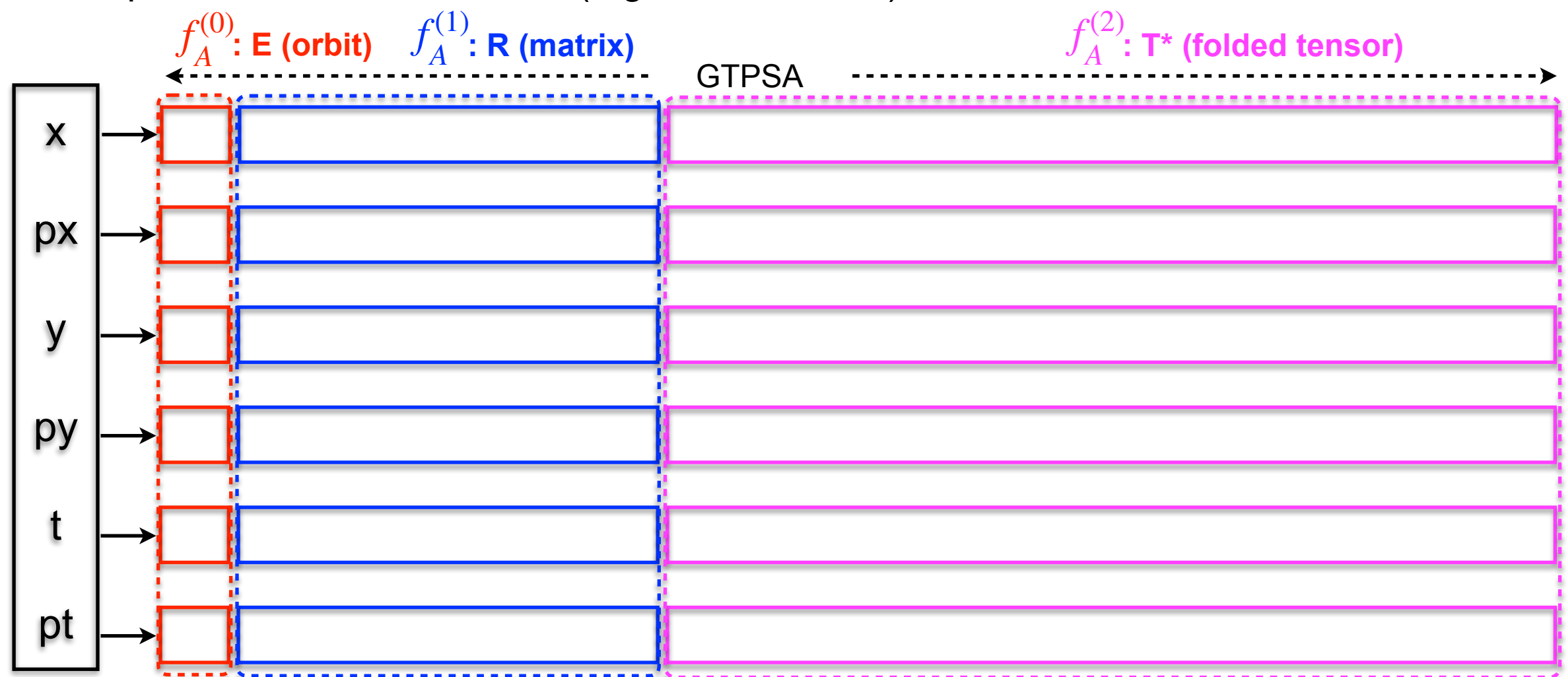
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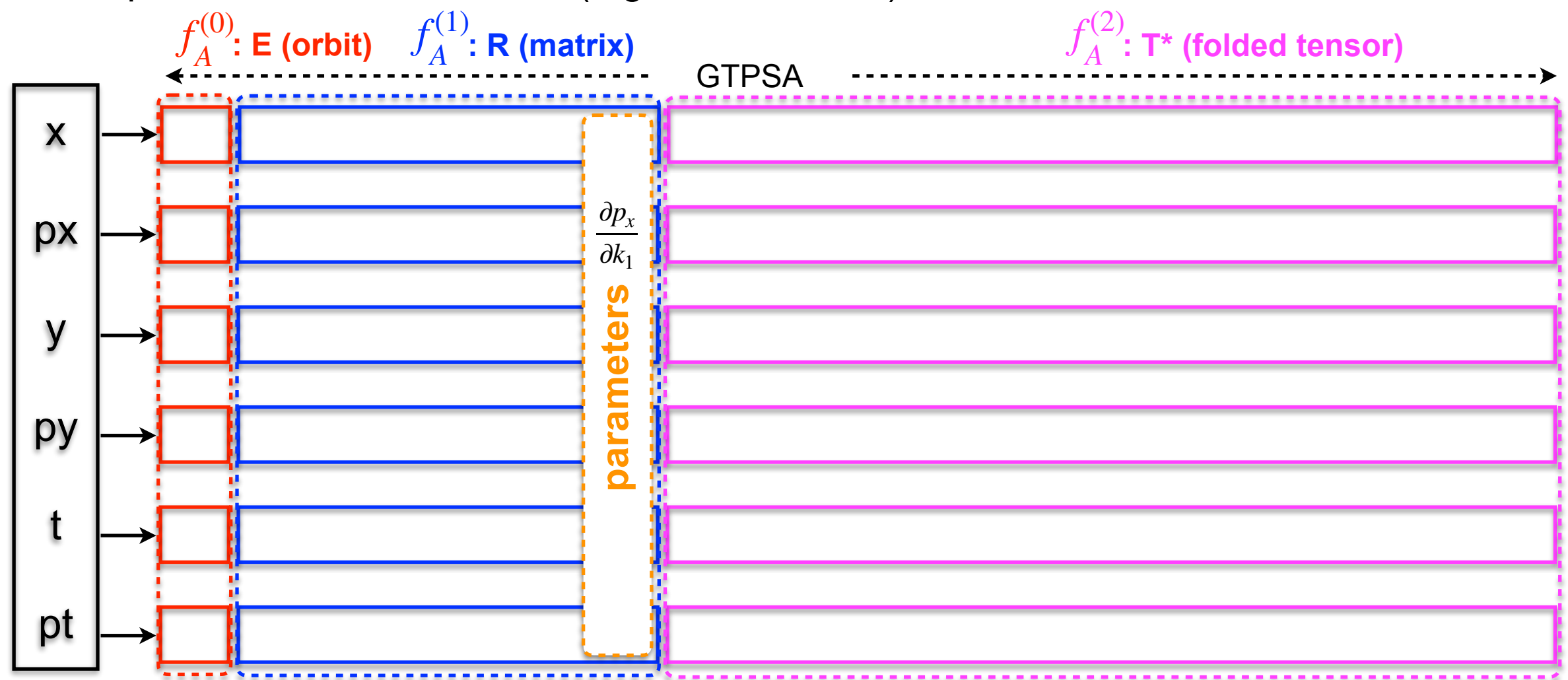
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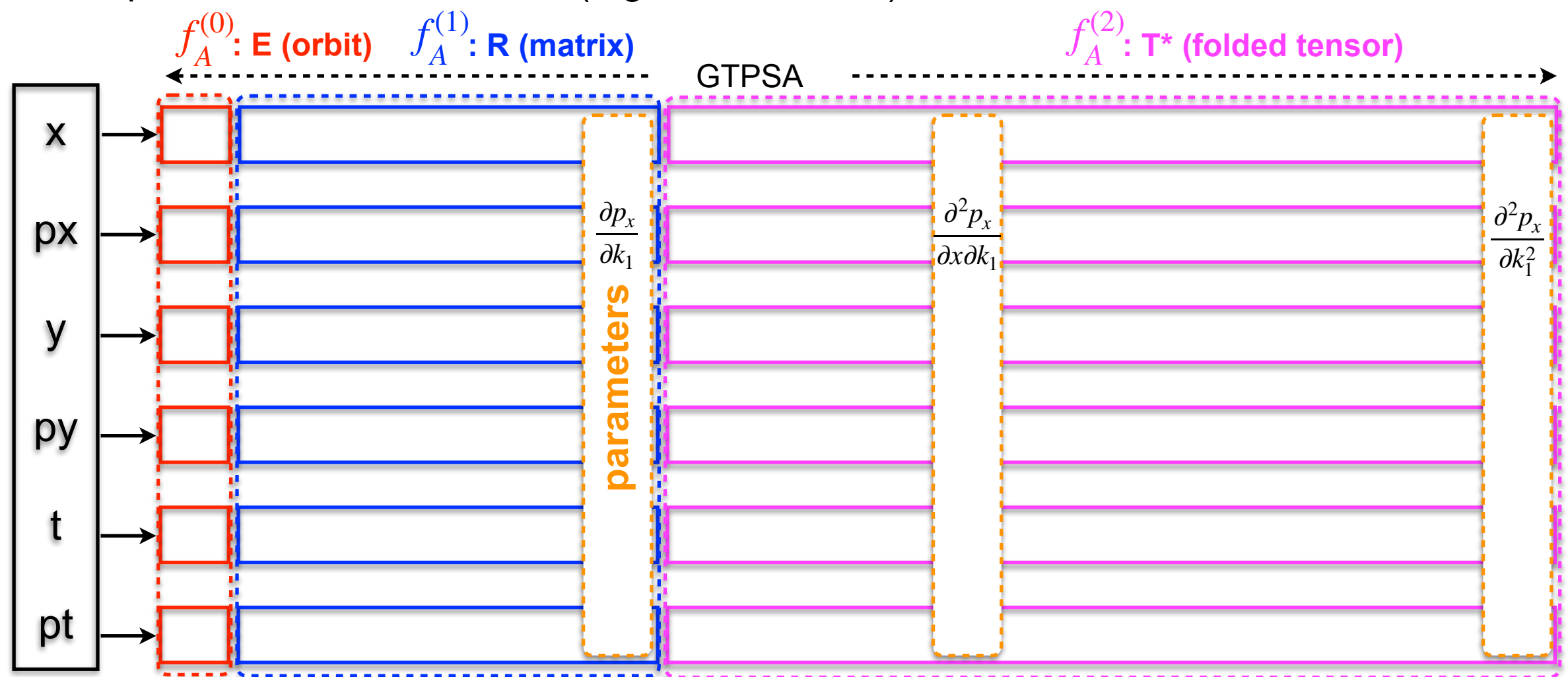
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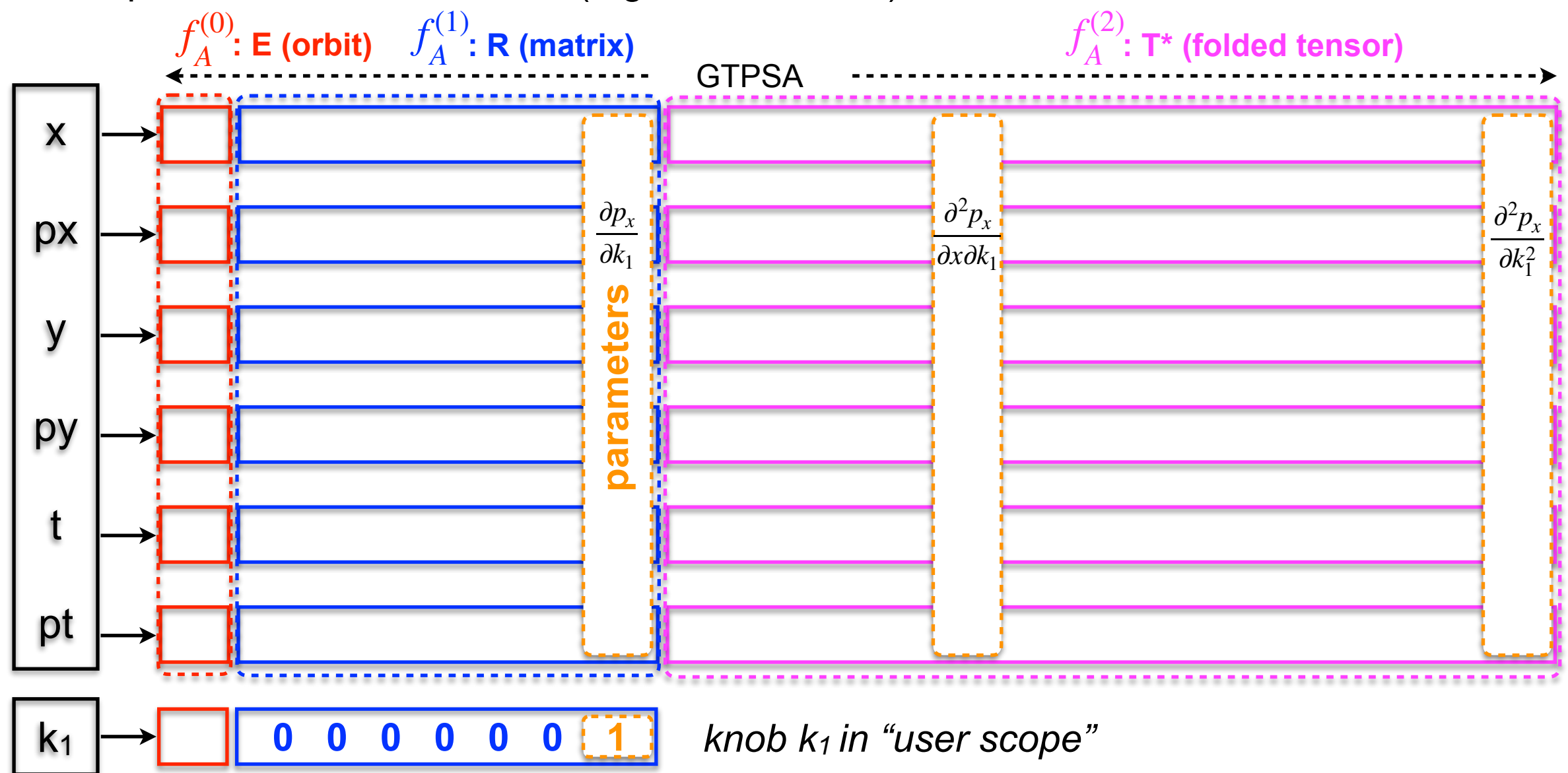
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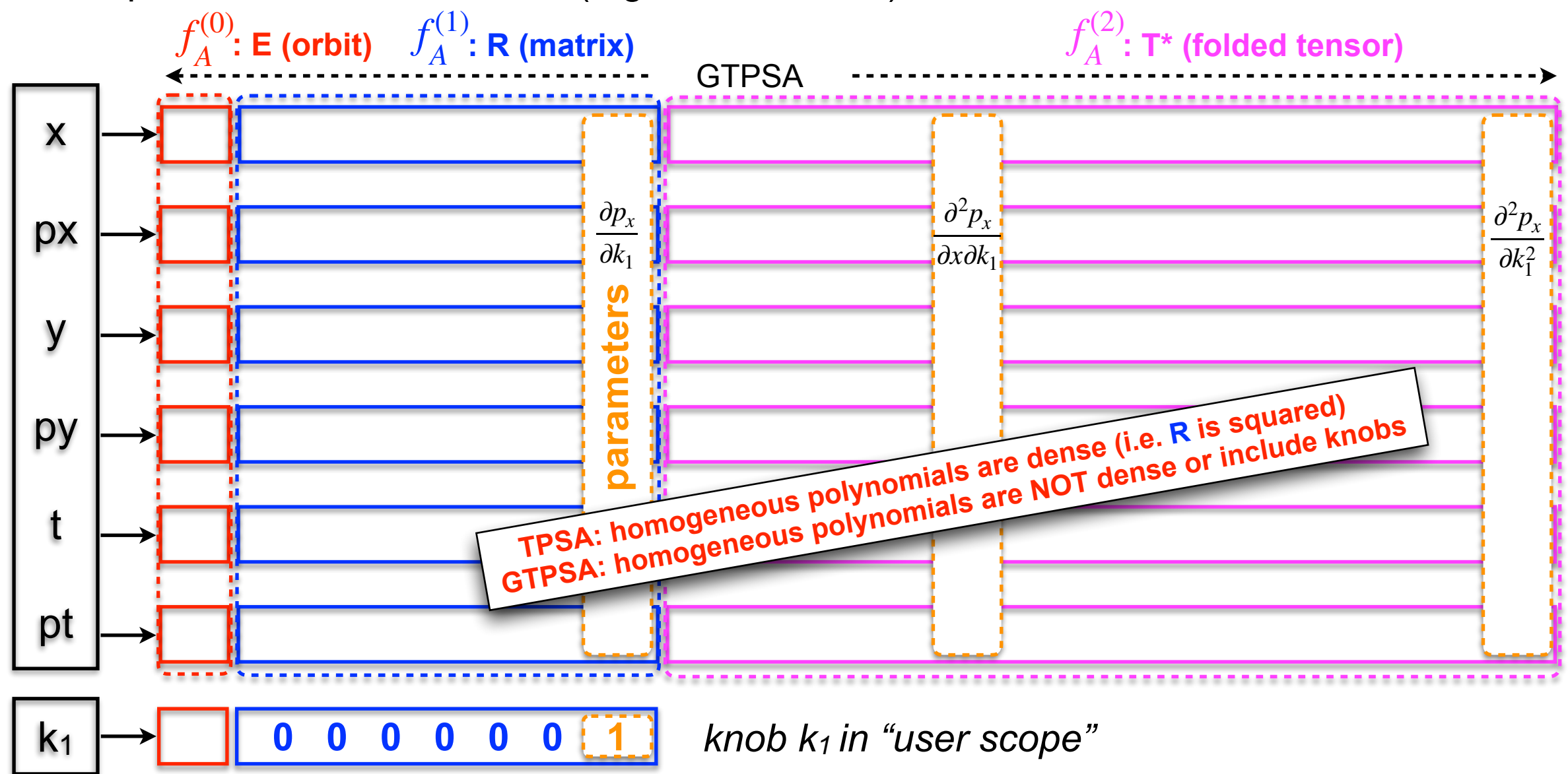
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DAmap vs. Matrix sizes

TPSA: homogeneous polynomials are dense with $\binom{n+v}{v} = \frac{(n+v)!}{n!v!}$ coefficients

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TPSA manipulate only numbers!

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TPSA are the only suitable solutions for high orders!

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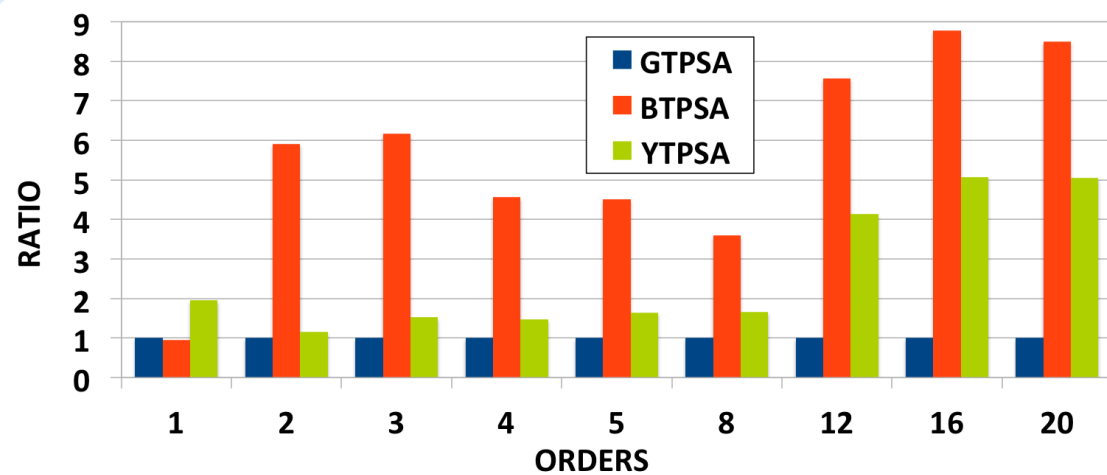


Fig. 5: Relative performance of the multiplications.

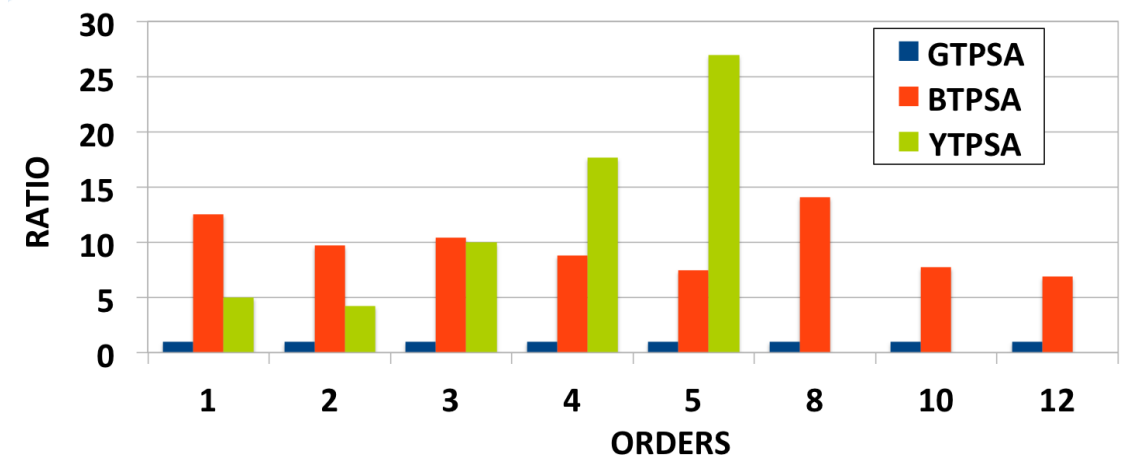


Fig. 6: Relative performance of the compositions.

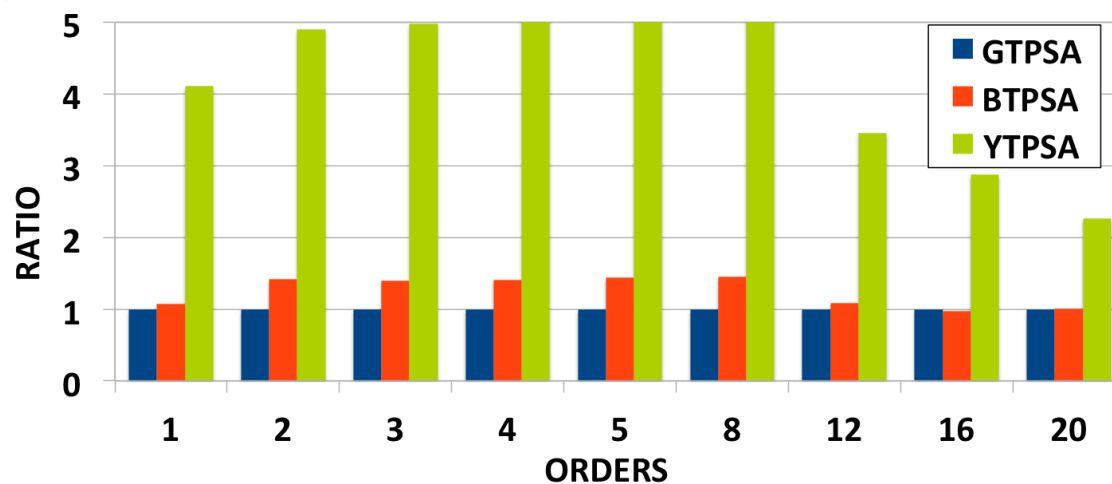


Fig. 2: Relative performance of indexing functions.

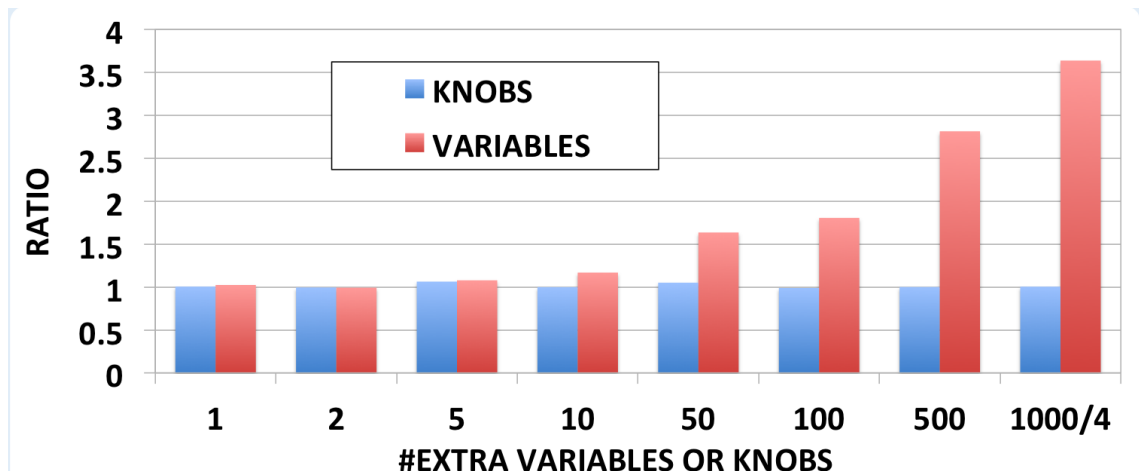


Fig. 4: Relative performance of multiplication at order 2 when using GTPSA with 6 variables and many knobs vs. homogeneous TPSA.

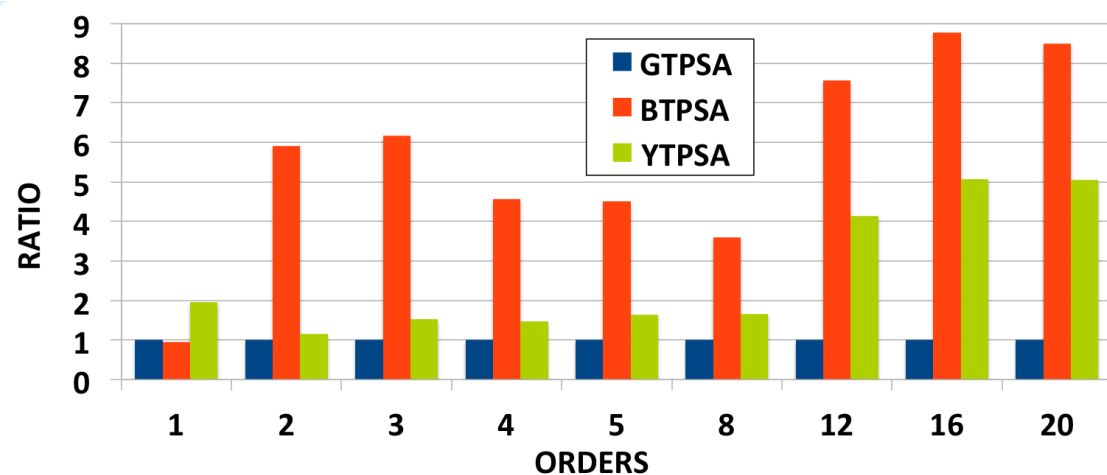


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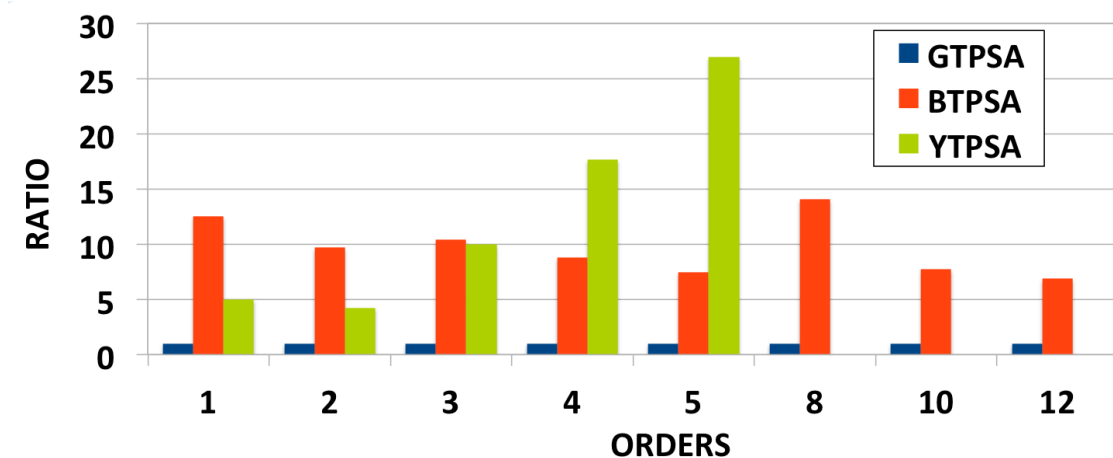


Fig. 6: Relative performance of the compositions.

The smaller the better!

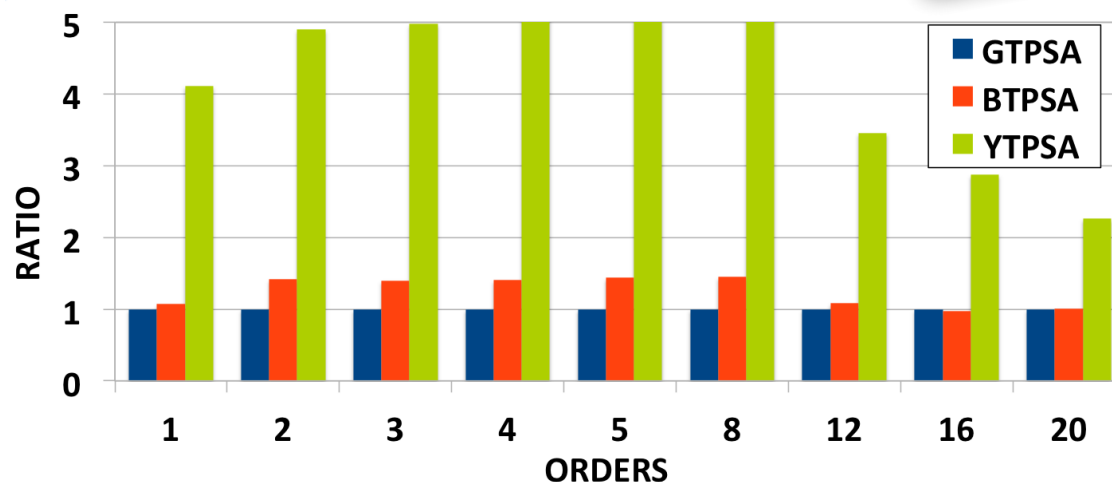


Fig. 2: Relative performance of indexing functions.

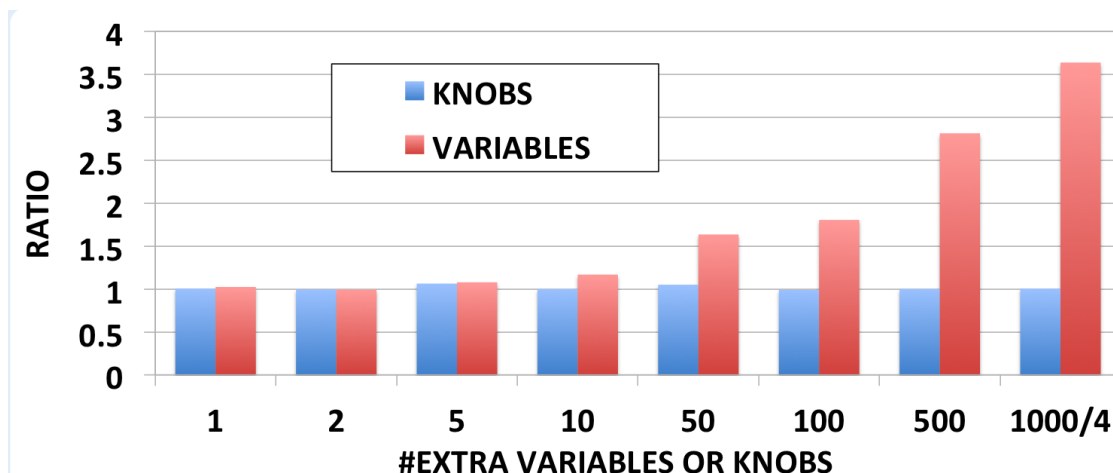


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